

# NexoSphere

Stock Market Sentiment Analysis and Data Retrieval Tool

#### **COEN 6313 PROJECT PRESENTATION**

**GROUP - #G15** 

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## OUTLINE

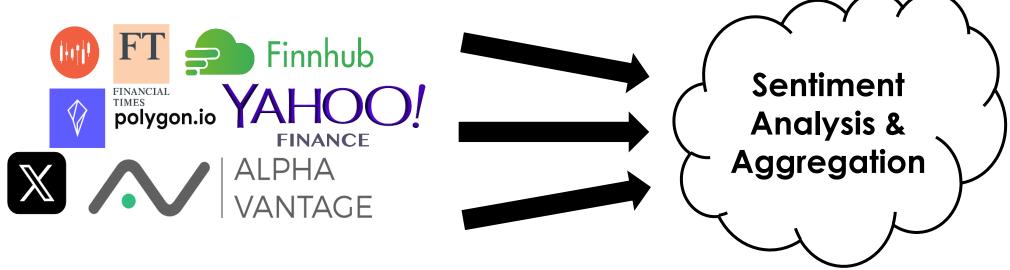
- 1) Functionalities
- 2) Architecture Design
- 3) Technical Choices
- 4) Data Model
- 5) Implementation

- 6) Project Schedule
- 7) Quality Attributes
- 8) Strengths & Limitations
- 9) Observations & Lessons Learnt
- 10) Member Contributions



### PROBLEM STATEMENT

- → Financial market participants struggle to efficiently access and analyze fragmented stock market data.
- → Investors lack tools to quickly gauge market sentiment from vast amounts of financial news.





## MAIN FUNCTIONS OF OUR SERVICE

#### Stock Market Data Retrieval and Sentiment Analysis

- Provision of Data Retrieval and Retrieval Service
- Provision of a Comprehensive Sentimental Al Analysis Tool

#### Stock Data Retrieval

- Retrieving **real-time** stock market data from **multiple** sources or **public API endpoints.**
- Combining these retrieved datasets into one object that **encompasses** all the data.

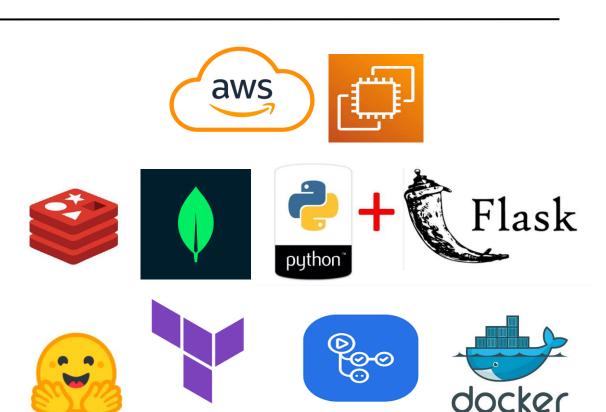
#### **Stock News Sentiment Analysis**

- Retrieving **real-time stock news and reports** from Financial News API.
- Performing **sentiment analysis** on the summarized content, using pre-trained Al model.



## TECH STACK

- Cloud Platform: Amazon Web Service (AWS) - EC2
- NoSQL Database: MongoDB, Redis
- Al Model Hosting: Hugging Face
- CI/CD VCS: GitHub, Terraform, GitHub Actions, Docker
- Programming Language: Python 3.10





HashiCorp

**Terraform** 



GitHub Actions





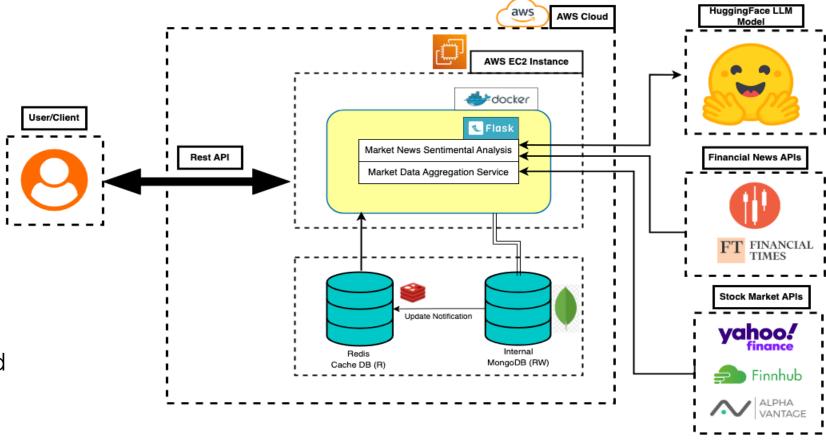
## ARCHITECTURE DESIGN

#### Monolithic Architecture

ConsolidatedDeployment

#### RESTful API:

- Statelessness
- Unified Interface
- Client-server based





## IMPLEMENTATION - Sentiment Analysis Service

Service: Provide summary and sentiment analysis to financial news

**Input:** HTTP GET request, specifying <u>ticker symbol</u>, <u>start\_date</u> and <u>end\_date</u> **Output:** <u>summarized news</u> and <u>sentiment analysis</u> for the stock over the given date range

#### **News Processing Pipeline:**

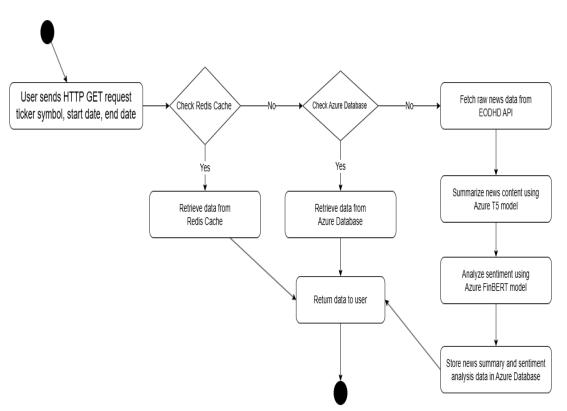
- News fetched from <u>public API</u>
- Summarized by Cloud T5 model
- Sentiment analyzed by Cloud FinBERT model

#### **Storage and Caching:**

- Processed data stored in (MongoDB)
- Frequently accessed data stored in Redis Cloud

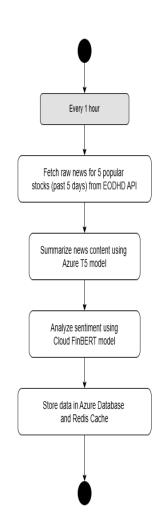


## IMPLEMENTATION - Sentiment Analysis Service



Activity Diagrams for Sentiment Analsysis workflows





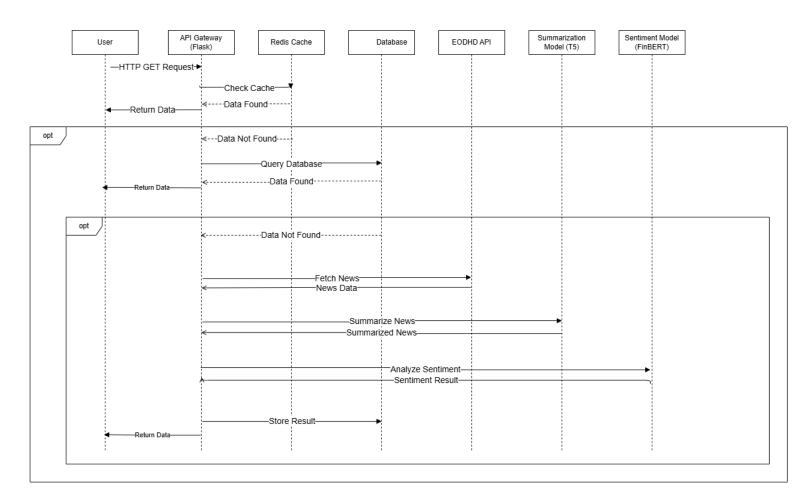
#### Data lookup and processing workflow:

- API gateway receives request
- 2. Lookup Redis Cache
- 3. Lookup <u>MongoDB</u>
- 4. News <u>Fetching and Processing</u>:
- 5. Fetches raw financial news from EODHD API
- 6. Summarizes and sentiment analyzes news
- 7. Store results to MongoDB, return result.

#### **Data Caching workflow:**

- Triggered by scheduler <u>hourly</u>.
- Read popular stock symbols from MongoDB
- 3. Fetches and Processes news for <u>last 5 days</u>
- Stores results to MongoDB
- 5. Caches results to Redis Cloud, <u>delete</u> outdated cache

## IMPLEMENTATION - Sentiment Analysis Service



#### <u>Sentiment Analysis Sequence Diagram</u>



#### **Core Classes:**

#### **NewsService:**

- 1. Handles database interactions.
- 2. Handles API calls
- Handles data processing

#### CacheNewsService:

- 1. Caching & Reading data
- 2. Managing outdated cache

#### **Data Management:**

#### MongoDB:

- 1. Stores news Labels, summary, sentiment analysis, news URL.
- Stores list of popular stocks for caching

#### **Redis Cache:**

- Stores data in JSON
- 2. Categorized by date

## DATA MODELS

```
"_id": {
    "$oid": "67298fad60e055ee3d46ccd4"
  "date": "2024-10-24T13:07:31+00:00",
 "title": "Tesla stock soars 13% thanks to an earnings beat and Elon Musk's 'best guesses'",
 "symbols":
   "1TSLA.MI",
   "TLO.BE",
    "TL0.F",
    "TLO.XETRA",
    "TL01.F"
    "TSLA.MX".
    "TSLA.NEO"
    "TSLA.US",
    "TSLA34.SA"
 "link": "https://www.yahoo.com/tech/tesla-stock-soars-13-thanks-130731296.html",
 "sentiment": {
    "label": "positive",
    "score": 0.9491851329803467
  "summarization": "Tesla (TSLA) stock is staying strong in trading Thursday after the company delivered earnings that largely
```

```
_id: ObjectId('67403562ff506e8058fd3da1')
    ticker: "TSLA"
    timeStamp: 2024-11-22T07:40:17.807+00:00

v data: Array (79)
v 0: Object
    timestamp: "2024-11-19 14:30:00"
    open: 335.43
v 1: Object
    timestamp: "2024-11-19 14:45:00"
    open: 338.39
v 2: Object
    timestamp: "2024-11-19 15:00:00"
    open: 337.46
```

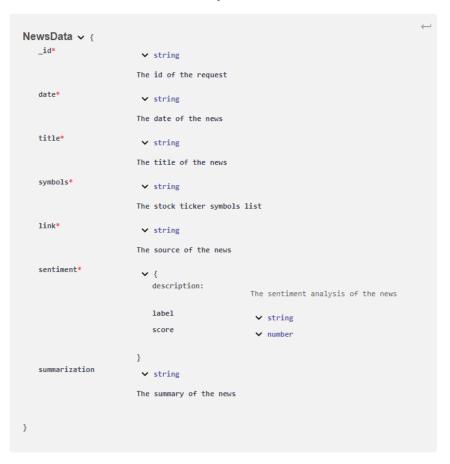
## Sentiment Analysis Data Model

#### Ticker Data Model

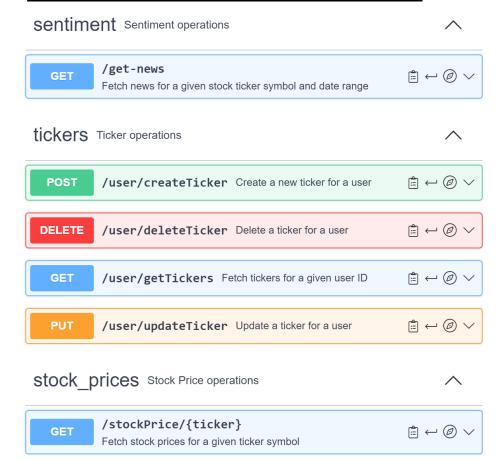


## IMPLEMENTATION — Swagger Hub (Extra Features)

#### <u>Sentiment Analysis Data Model</u>



#### RESTful API Listing of the Routes

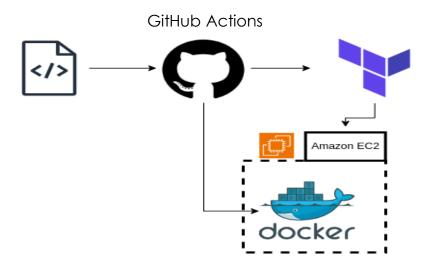


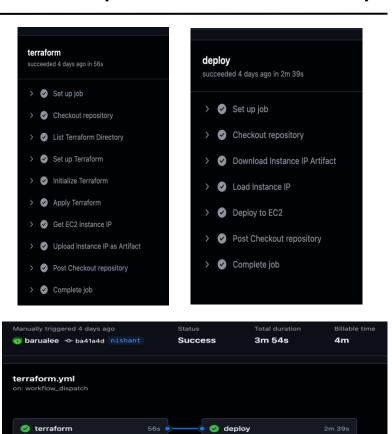


## IMPLEMENTATION — CI/CD Deployment (Extra Features)

## To Implement CI/CD, we have used the following workflow using GitHub Actions:

- Checkout source code
- Terraform to instantiate our EC2 cloud instance on AWS
- Deploy and run the docker container







## QUALITY ATTRIBUTES

- Availability: AWS's global network of data centres ensures high availability and low-latency
  access to users across different regions. Similarly, MongoDB ensures high availability through
  automatic replication and failover, distributing data across multiple nodes to prevent a
  single point of failure.
- **Performance:** We are leveraging Redis as a caching solution on top of our MongoDB solution to store our analysis results, which improves performance.
- **Scalability:** AWS EC2 offers both vertical and horizontal scaling, enabling rapid adjustment of resources to match varying workload demands. MongoDB supports horizontal scaling through sharding, distributing data across multiple servers as it grows. Docker containers can be easily scaled using orchestration tools like Kubernetes, allowing applications to scale seamlessly based on demand.



## OBSERVATIONS & LESSONS LEARNT

- **Data Acquisition and Quality**: Unavailability of free/affordable real-time data sources that are free of request and access limits.
- **Model Complexity and Performance**: Choosing the right model architecture and hyperparameters can be complex.
- **CI/CD Deployment:** Automated deployment using CI/CD, utilize GitHub secrets to securely store environment variables and ssh keys.
- **Al Model:** Consciously terminate the launched Hugging Face Al models to avoid incurring extra charges.
- Lack of User Authentication: All incoming requests without user authentication, potentially leading to unauthorized access and system overload.
- **Data Processing Delay**: Accessing sentiment analysis for less popular stocks can result in longer processing times.



## STRENGTHS & LIMITATIONS

- Flexible Data Models: MongoDB's document-based model allows for dynamic, flexible schema changes, making it ideal for applications dealing with semi-structured or evolving data.
- Powerful Querying and Analytics: MongoDB provides advanced querying and aggregation pipelines.
- **High-Performance Caching**: Redis operates as an in-memory database, providing fast data access by storing information in memory; highly effective for caching.
- **Pre-Trained AI Models**: Hugging Face offers pre-trained AI models, improving development time and eliminating the need for training from scratch.
- Collaboration and Automation: GitHub provides version control and collaboration features. GitHub Actions
  automates CI/CD pipelines, enabling seamless workflows for building and deploying applications. Docker offers
  portability and consistency.
- Memory Constraint: Redis, being in-memory, is limited by available RAM and managing growing data can be challenging without proper management policies.
- Deployment Costs: Hosting large AI models on Hugging Face are costly, particularly for real-time applications.
- Latency: Hugging Face models experience latency in real-time, under high-volume traffic or models requiring high compute power.



## SCHEDULE & PROJECT MANAGEMENT

GANTT CHART NEXOSPHERE								
TASKS	SEPTEMBER	OCTOBER	NOVEMBER					
PROBLEM STATEMENT								
TECHNOLOGY CHOICE								
SENTIMENT ANALYSIS								
AI MODEL INTEGRATION								
AGGREGATION & FRONTEND								
CI/CD								
SWAGGERHUB								



## MEMBER CONTRIBUTIONS

Task List	40188139 (Ismael B.)	40270361 (Zihan M.)	40280162 (James A.)	40267821 (Nishant B.)	Total
Problem Description	25%	20%	30%	25%	100%
Tech Stack & Review	25%	20%	30%	25%	100%
Data Model	20%	20%	30%	30%	100%
Sentiment Analysis Service	10%	65%	15%	10%	100%
Al Model Integration	10%	50%	30%	10%	100%
Data Aggregation Service	60%	10%	15%	15%	100%
Frontend	25%	10%	25%	40%	100%
CI/CD Deployment	35%	5%	10%	50%	100%
Reporting & Documentation	15%	25%	40%	20%	100%



## Links

- https://github.com/NexoSphere-Org/NexoSphere
- http://ec2-35-153-83-207.compute-1.amazonaws.com:8080/
- http://ec2-35-153-83-207.compute-1.amazonaws.com:8081/getnews?ticker\_symbol=MSFT&date\_start=2024-11-20&date\_end=2024-11-25
- <a href="http://ec2-35-153-83-207.compute-1.amazonaws.com:8081/user/getTickers?userId=12347">http://ec2-35-153-83-207.compute-1.amazonaws.com:8081/user/getTickers?userId=12347</a>
- http://ec2-35-153-83-207.compute-1.amazonaws.com:8081/stockPrice/TSLA?start\_date=2024-11-17&end\_date=2024-11-21
- https://app.swaggerhub.com/apis/ISMAELRIDHA/nexosphere\_api/1.0-oas3



# THANK YOU

# NexoSphere

