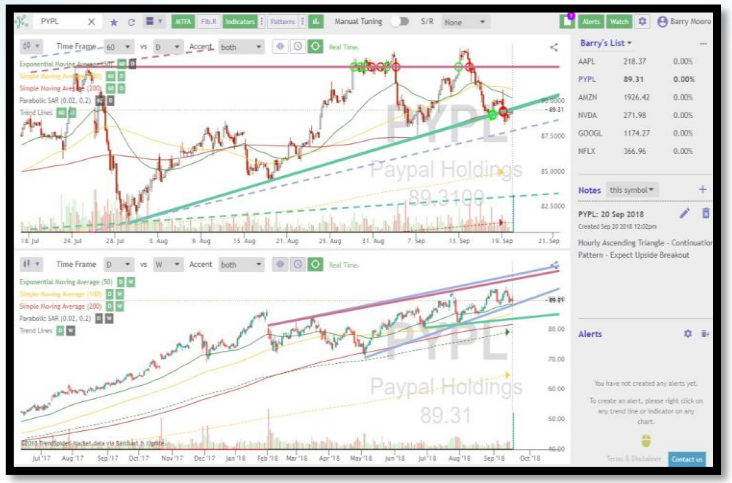
Trade For Me

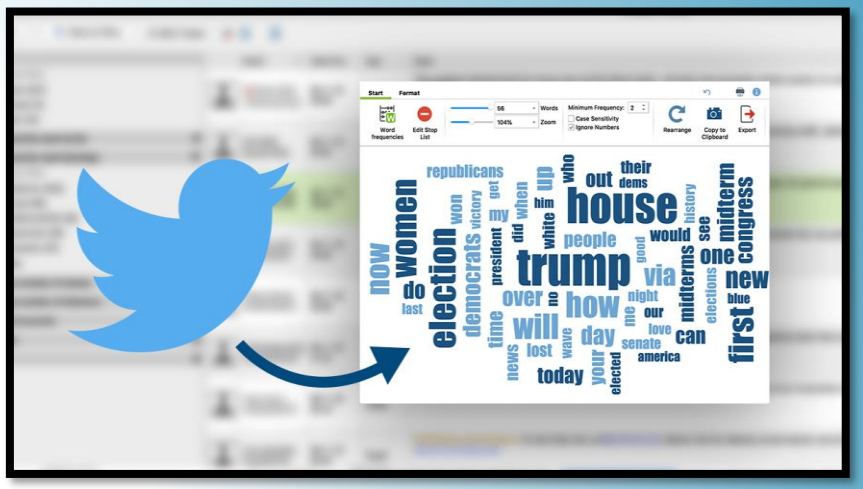
Srikanth Polisetty
U7142680

Supervisors

Dr Bernardo Pereira Nunes, Dr Artem Lenskiy

Overview





- Often it is very confusing to take a decision on how and when to trade shares
- Trading consultants use human expertise and ML based algorithms for decision making
- Twitter sentiment analysis allows us to predict possible trade options
- This project TFM (Trade For Me)
  - Calculates the reliability of twitteratis based on correlation between their tweets and actual variations in share prices
  - Suggests investors in trading decisions by considering the latest tweets from the reliable tweeters

TFM overall design

- The TFM GUI maintains two level of users (Admin, customer)
- **Admin** type user can
  - Create / Delete / Modify users
  - Create / Delete / Modify sectors and companies
  - Do sentiment analysis and identifies reliable twitteratis
- **Customer** type user can
  - specify the investment amount and request for trading suggestions
  - Display the companies on which the investment can be made based on recent tweets of reliable twitteratis
  - Place the order for the trading

TFM critical operations …1/2

When Admin requests to **Get Reliable Twitteratis** (Ref 3.3) for a given sector, program will

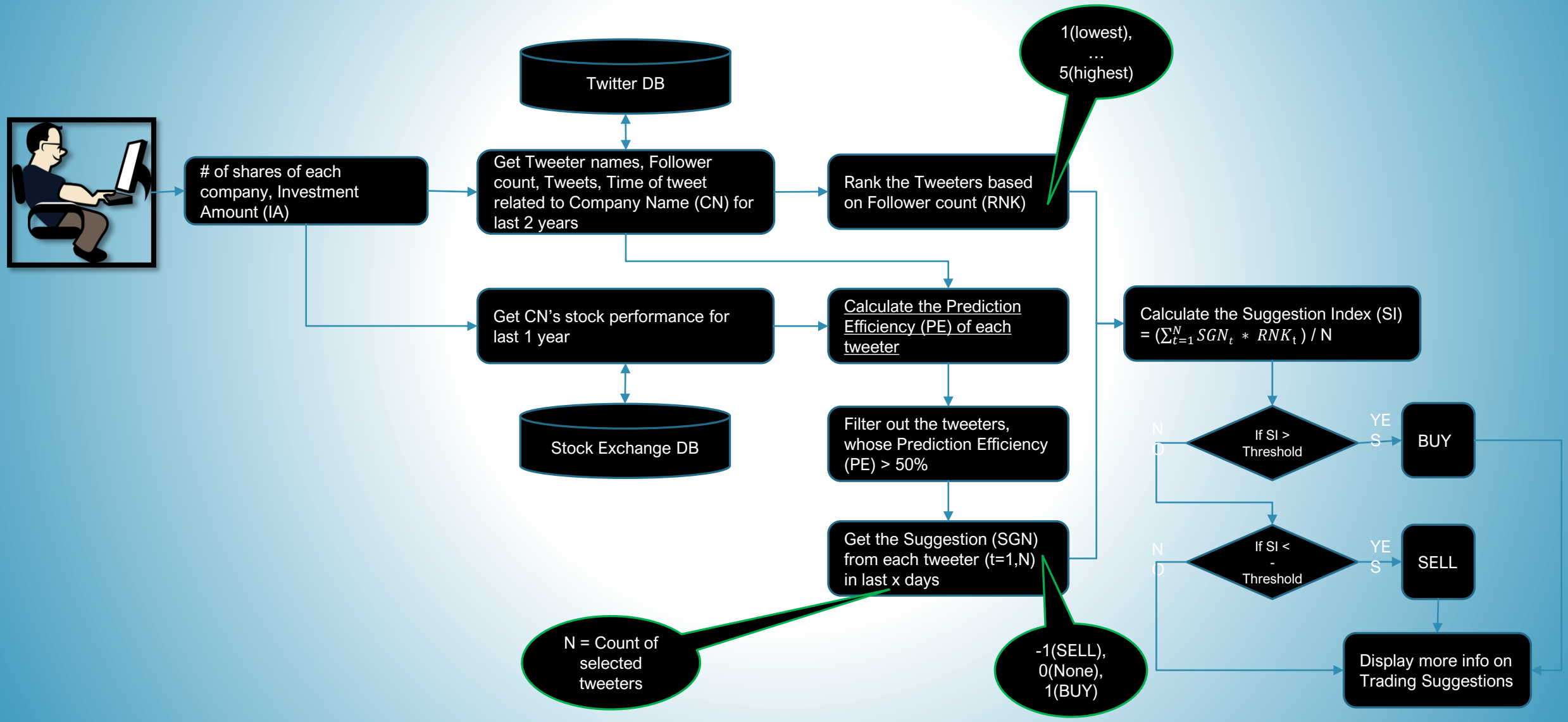
1. Fetch all the company names for specified sector and display
2. Fetch all the tweets for specified duration
3. Filter the tweets relevant for specified company or for companies of specified sector
4. Fetch the stock price historical data for associated companies of given sector
5. For each of the day within specified duration, correlation factor between tweets from a given twitterati and share price variation shall be calculated.
6. The average of correlation factors for all the days within specified duration shall be stored in Correlation table for each Twitterati and for each company.

TFM critical operations …2/2

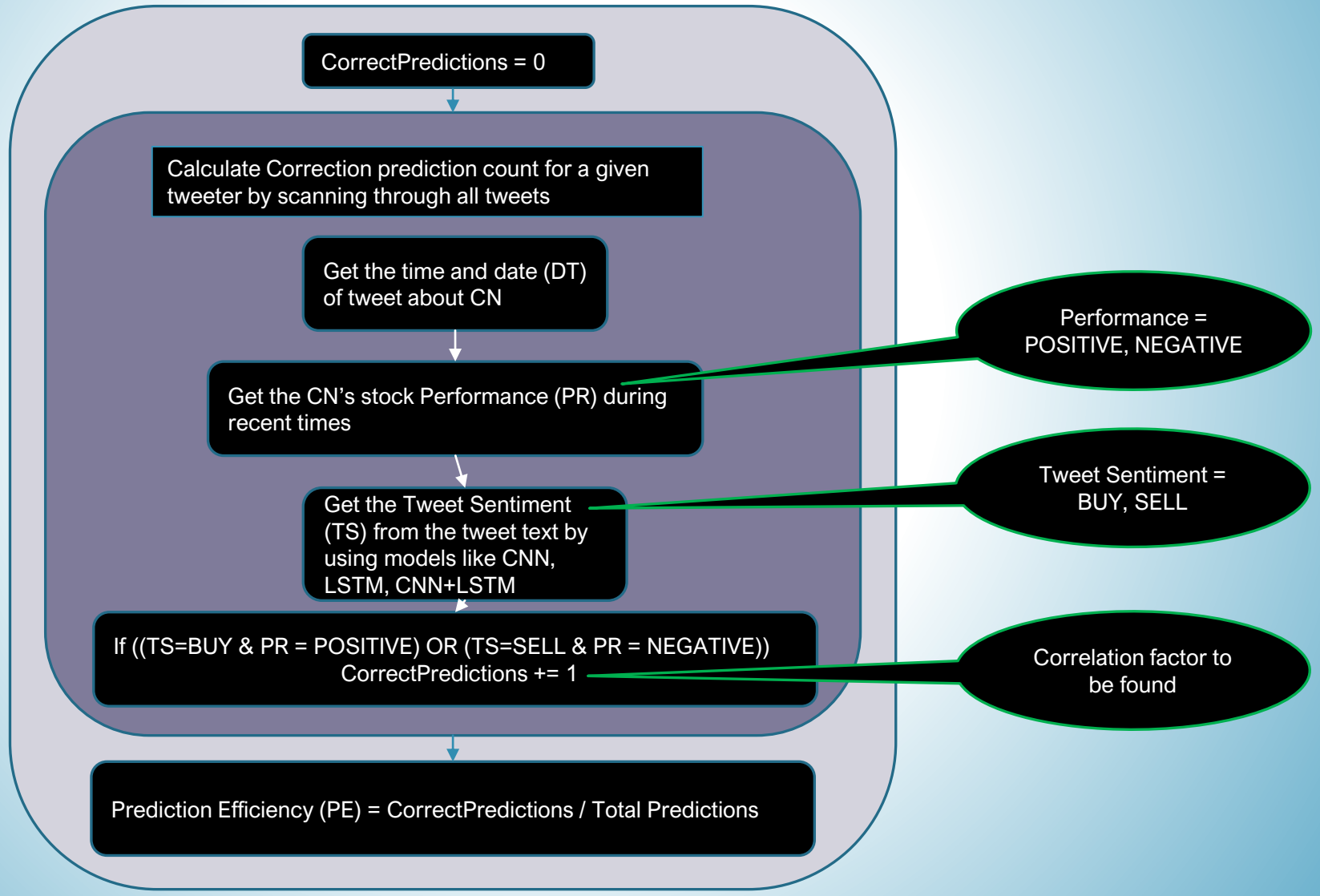
When Customer requests to **Get Trading Suggestions** (Ref 3.4) for a given sector, program will

1. Get the investment (USD) available from the customer.
2. Fetch the Correlation entries from Correlation table for specified sector.
3. If a company has more than one reliable Twitterati, the entry corresponding to Maximum Correlation_Factor shall be chosen.
4. Depending on the investment limit and current share price of the chosen companies, displays the companies for trading
5. Customer can specify the number of shares to be purchased for each of those companies and submit the form.
6. TFM system shall submit the purchase request if the investment available is sufficient

Architecture Diagram



Prediction Efficiency Calculation

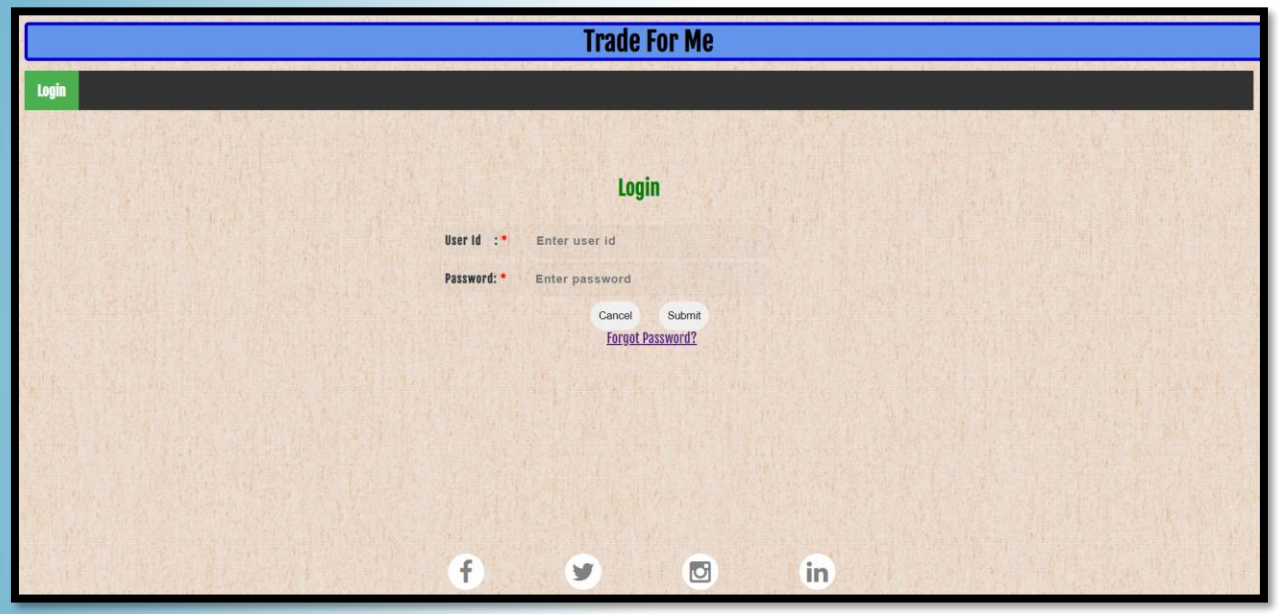


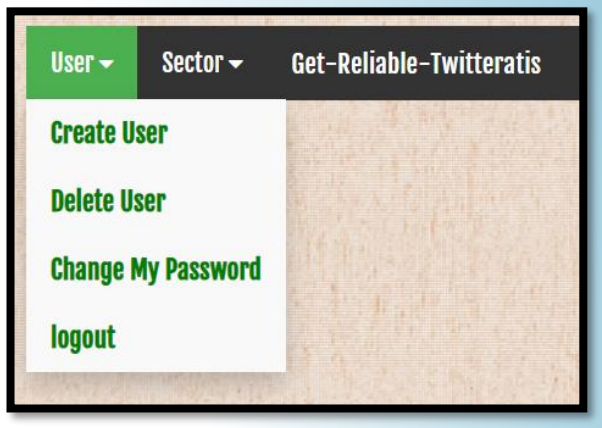
Database design

Database Tables

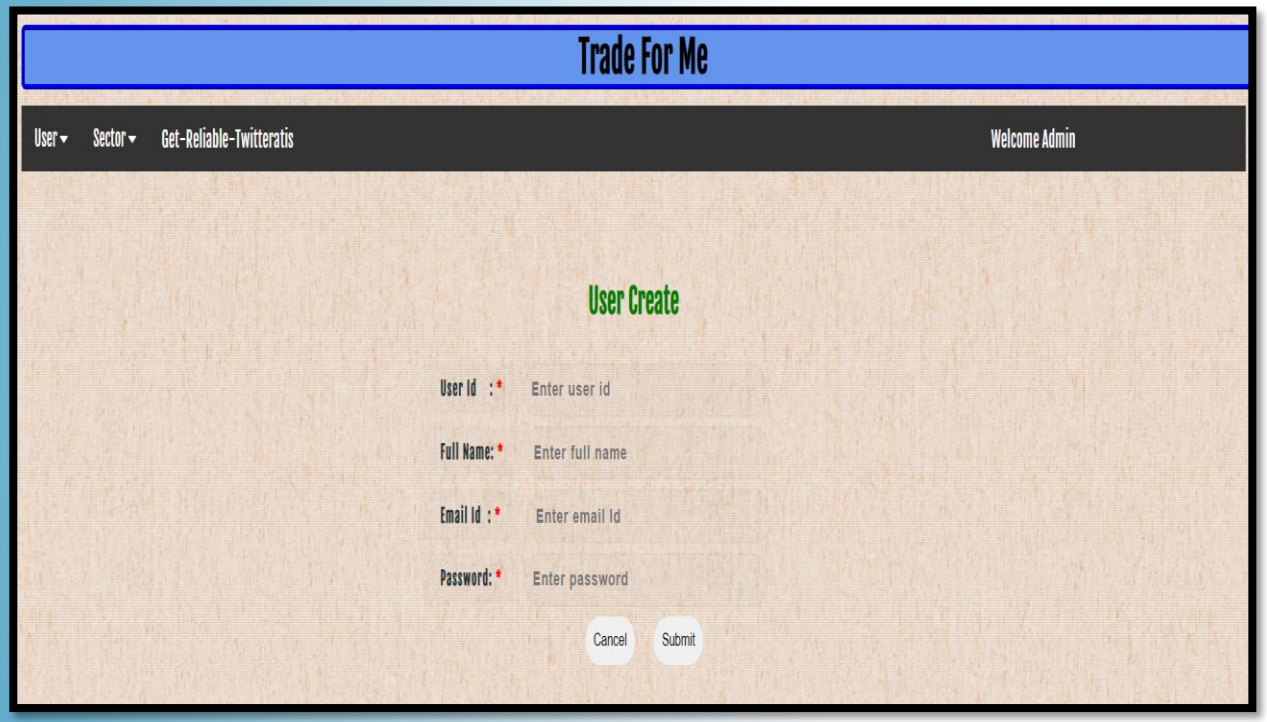
- Twitterati Info
- Twitter data Info
- Sector Info
- Company Info
- Stock Data
- Correlation Data

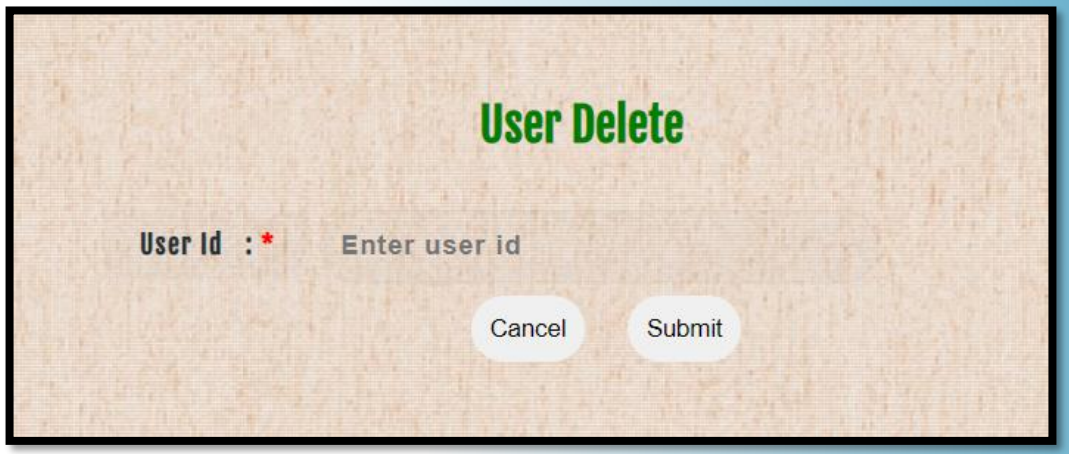
User interface: Login

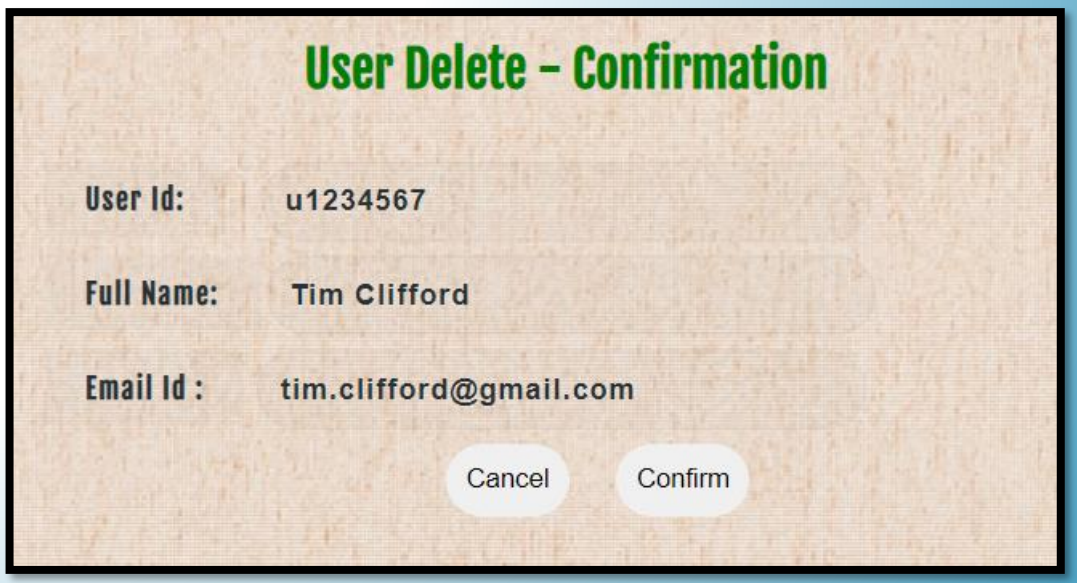




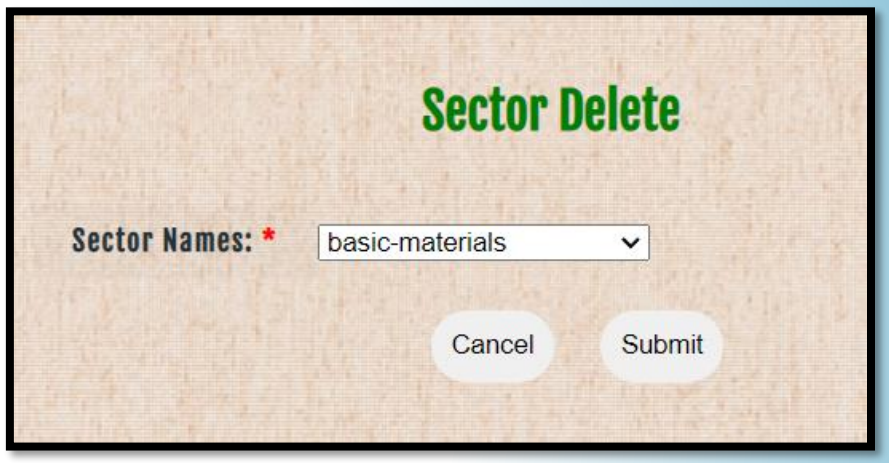
User interface: User management





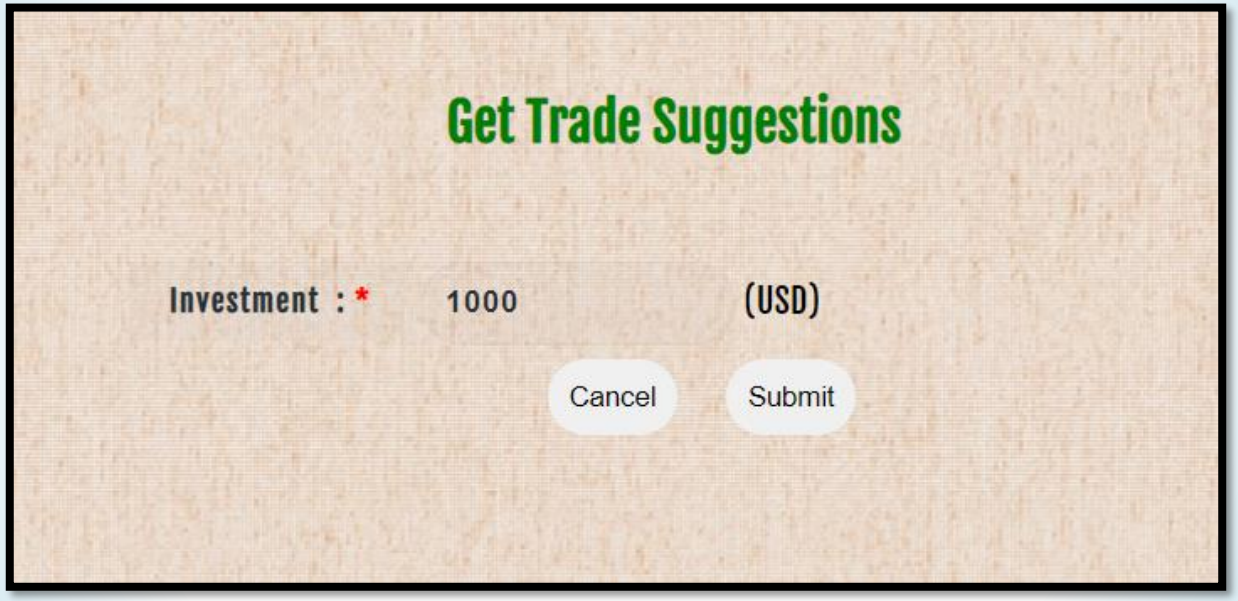


User interface: Sector management

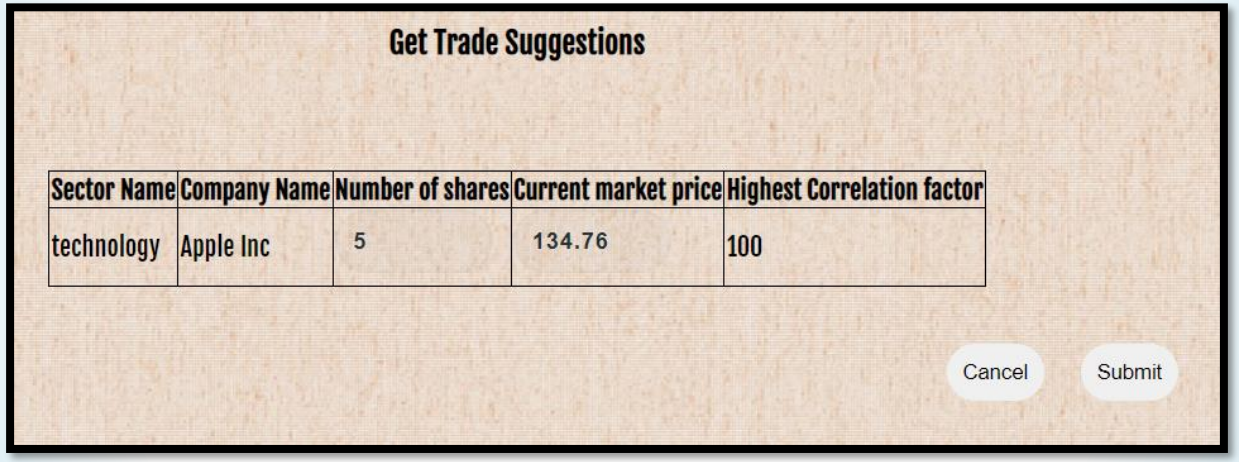


User interface: Get reliable twitteratis

User interface – Investment specification



User interface: Trade suggestions



Get Trade Suggestions

Sector Name	Company Name	Number of shares	Current market price	Highest Correlation factor
technology	Apple Inc	5	134.76	100

Cancel Submit

User interface: Trade confirmation

Future extensions

- To improve the precision of the strategy, more complex algorithms such as Restnet and VGG16 can be employed. Text classification is a possible area for improvement. Here, the strategy may result in the temporary loss of a user. Hence, there is need for improvement.
- When there are no tweets about a company from users with high score in recent times, the previous prices of shares of that company are used to make predictions about the future share prices. Various models like LSTM, RNN, CNN, BiLSTM, Arima are used for time series forecasting. A bar plot comparing the root mean squared error of those models is as follows.

Acknowledgements

- I would like to express my heartfelt gratitude to my professors (Dr Bernardo Pereira Nunes, Dr Artem Lenskiy) for their invaluable guidance and support throughout the project. Their expertise and willingness to share their knowledge have been instrumental in helping me understand the intricacies of the research subjects and making this project a success.
- I am particularly grateful for their guidance in providing me with important resources such as research papers and other literature that have been extremely helpful in understanding the subject matter and building the foundation for my project. Their constant encouragement and support have been a source of motivation and inspiration throughout the project.

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Thank you