

### GROUP 29

# Analyzing Social Media Posts for Mental Health Disorder Detection

ROLL NUMBER	NAME	
13000121033	SOUMYADEEP NANDY	
13000121037	PRITHWISH SARKAR	
13000121040	SAGNIK MUKHOPADHYAY	
13000121058	ARKAPRATIM GHOSH	

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PROJECT-III ( PROJ-CS881 )

CSE : SEMESTER 8
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#### **MOTIVATION**

- → Rising global concern over mental health disorders: Mental health issues are affecting millions worldwide, requiring urgent attention .
- → Social media is a key outlet for emotional expression : Platforms like Twitter and Reddit reveal mental health struggles in real-time.
- → Early detection of mental health issues can save lives : Identifying mental health disorders early helps provide timely interventions.
- → Machine learning can automate detection of mental health disorders :

  Technology enables efficient analysis of large social media data for early warning signs.
- → Potential to assist mental health professionals and organizations :

  Provides valuable insights for mental health monitoring and public health efforts.
- → Opportunity to improve mental health awareness on social platforms

  Can support campaigns that foster awareness and reduce stigma online.

### INTRODUCTION

- Role of social media in mental health expression: People share emotions, struggles, and experiences on platforms like Twitter and Reddit.
- → Goal of the project: To detect mental health disorders early through the analysis of social media posts (text and image) or by uploading images, videos, facial expression recognition, generating image captions , manually inserting text and retrain model in the process.
- → Leveraging machine learning: Using advanced techniques like NLP and classification models to analyze text normally posted, extracted from image, audio and video.

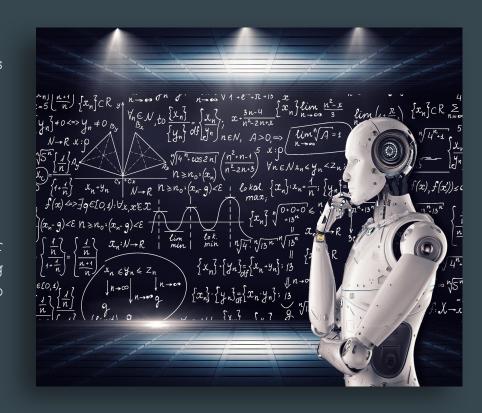


### INTRODUCTION (CONTINUED)

- → Focus on text classification for the base model : Analyzing language patterns to classify posts related to mental health issues.
- → Impact of early detection : Can enable timely intervention and direct users to mental health support services.
- → Models used in the project:

  Techniques like Logistic

  Regression, XGBoost are applied for high accuracy. Ensemble Model using the two former models is created to get higher accuracy
- → Broader goal : Use technology to assist mental health professionals and enhance public health awareness.



### **RESEARCH WORK**

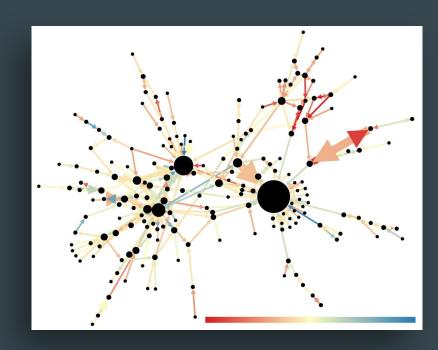
→ Social media and mental health research

Explored studies on how social media data can reveal mental health conditions.

- → Key study by Choudhury et al. (2013)
  Showed the predictive power of Twitter data in identifying depression through linguistic patterns.
- → Guntuku et al. (2017) review

  Synthesized various approaches to detecting mental illness using sentiment analysis on social platforms.
- → Mathur et al. (2022) systematic review

Highlighted the success of machine learning techniques in detecting disorders like depression and anxiety.



### **RESEARCH WORK (CONTINUED)**

- → Nadeem (2016) study on Twitter

  Demonstrated the potential of text

  analysis to identify at-risk

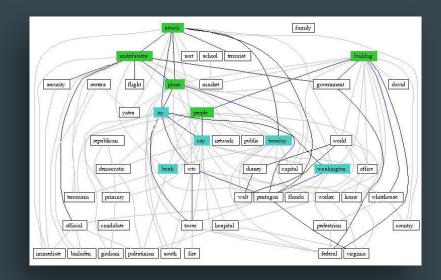
  individuals based on emotional cues in

  tweets.
- → Al Sagri and Ykhlef (2020) approach
  Combined linguistic and behavioral
  features for more accurate depression
  detection on Twitter.
- → Recent study by Vaishnavi et al. (2022)

Comparative analysis of algorithms to identify mental health conditions from social media posts.

→ Ethical considerations by Safa et al. (2023)

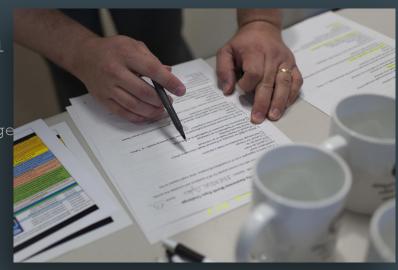
Addressed data privacy challenges in mental health detection research using social media data.



### PROBLEM DEFINITION

- → Rising prevalence of mental health disorders: Increasing cases of depression, anxiety, PTSD, and other mental health issues globally.
- → Challenges in early detection : Mental health problems are often undiagnosed until advanced stages, limiting timely intervention.
- → Vast amount of unstructured social media data: Social media platforms generate large volumes of text that can indicate mental health struggles.
- → Need for efficient detection methods :

  Manual analysis of social media posts is
  time-consuming; automation using machine
  learning is essential.
- → Goal: To develop a system that accurately classifies social media posts based on mental health disorders.



### PROPOSED WORKFLOW

#### → Data Collection

Collect Reddit posts using the PRAW API and extract labeled data for mental health categories like Normal, Anxiety, Depression, PTSD, and Bipolar.

#### → Data Preprocessing

Clean the collected data by removing URLs, stopwords, and special characters. Normalize and tokenize the text, followed by converting it into numerical formats ( TF-IDF, BagOfWords ) for analysis.

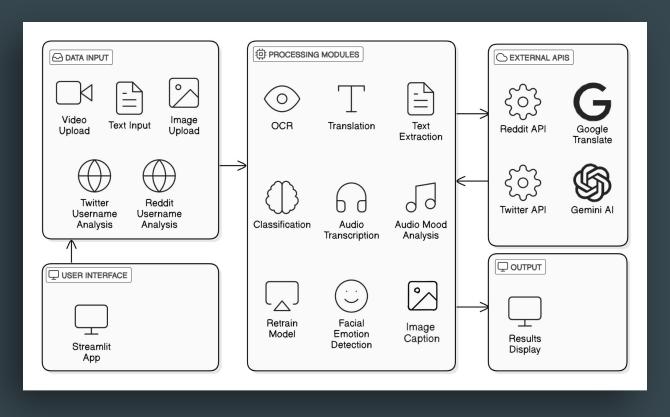
#### → Model Training and Evaluation

Train machine learning models including Logistic Regression, Naive Bayes, SVM, Random Forest, XGBoost, KNN, LSTM and Transformers. Evaluate their performance using metrics such as accuracy, precision, recall, and F1-score. Apply Hyperparameter Tuning on ML models as needed to improve the accuracy further. Leverage Ensemble Learning with multiple models to get higher accuracy.

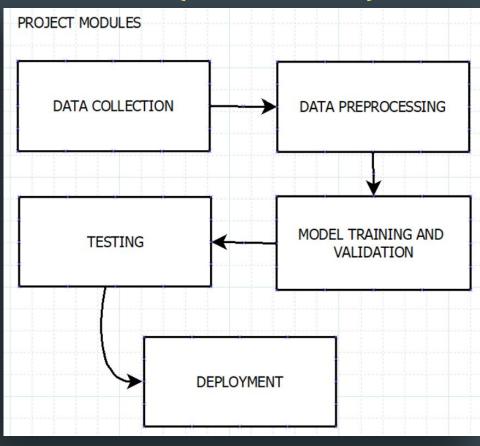
#### → Testing and Deployment

Test the best-performing models and deploy them on a user-friendly interface using Streamlit. Ensure the system supports real-time classification for various inputs.like text, image, video, audio and user profiles from social media platforms like Reddit and Twitter.

## PROPOSED WORKFLOW (CONTINUED)



## PROPOSED WORKFLOW (CONTINUED)



#### **IMPLEMENTATION**

- → Data Collection: Scraped Reddit posts related to mental health issues and for analysis.
- → Data Preprocessing: Cleaned and normalized the text by removing URLs, stop-words, punctuation, and applied tokenization and lemmatization techniques.
- Feature Extraction: Utilized Bag of Words, Term Frequency-Inverse Document Frequency (TF-IDF) to convert text into numerical format for machine learning models.
- → Splitting the Dataset: Divided the dataset into training and testing sets to train models and evaluate their performance.

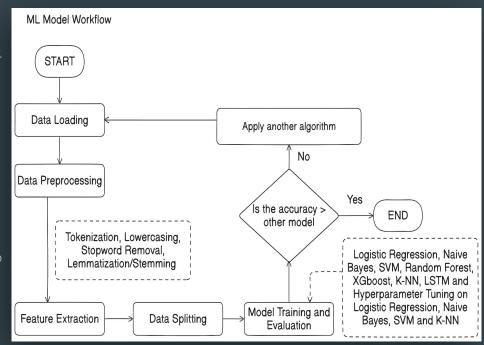
Rqmt =	Requirement Item =	Requiremen t Analysis = Status
FR-001	Collect social media data from Reddit.	Completed ▼
FR-002	Implement data cleaning and preprocessing.	Completed -
FR-003	Train machine learning and deep learning models.	Completed •
FR-004	Evaluate models using performance metrics (accuracy, recall, F1 Score, Support).	Completed •
NFR-001	Testing different features of the web application	Completed •
NFR-002	Final Web Application Deployment	Completed •

FUNCTIONAL REQUIREMENTS

### **IMPLEMENTATIONS (CONTINUED)**

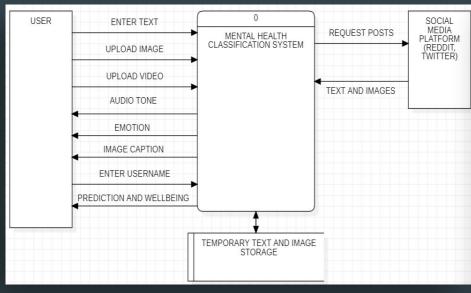
- → Model Training: Trained multiple models: Logistic Regression, k-Nearest Neighbors (k-NN), Support Vector Machine (SVM), Naive Bayes, Random Forest, XGBoost, Long Short Term Memory and Ensemble Model
- → Hyperparameter Tuning: Applied RandomizedSearchCV to optimize the performance of each model.
- → Model Evaluation: Used metrics like accuracy, precision, recall, F1-score, and confusion matrices to evaluate model effectiveness.
- → Prediction and Deployment:

  Implemented the best-performing model for predicting mental health issues from social media posts and deployed using Google Colab.

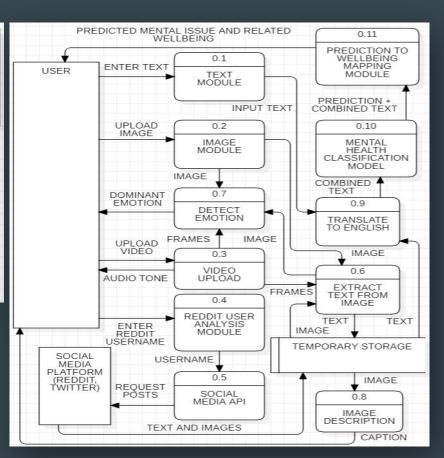


MODEL WORKFLOW

### **IMPLEMENTATION (CONTINUED)**



DFD LEVEL 0



DFD LEVEL

### **RESULTS AND ANALYSIS**

#### → Logistic Regression

Achieved consistent performance with highest accuracy. Precision and recall indicate reliable classification for balanced datasets.

#### Naive Bayes

Performed well with text data, especially for independent features. Precision was slightly lower for imbalanced classes but remained effective overall.

#### Support Vector Machine (SVM)

Delivered high accuracy for nonlinear classification tasks. The model showed robustness with complex feature interactions.

#### Random Forest

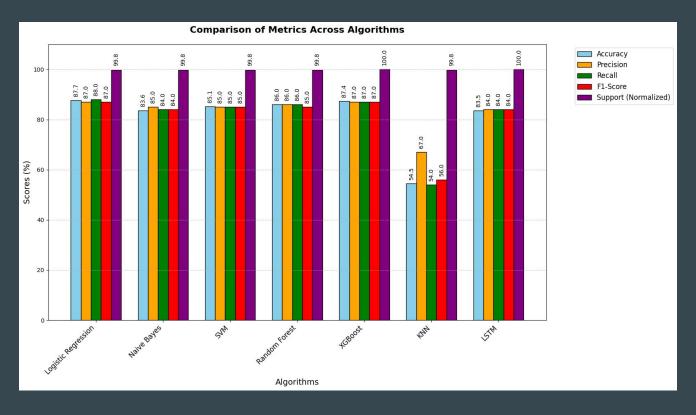
Provided strong classification performance with reduced overfitting. Achieved balanced precision and recall across all mental health classes.

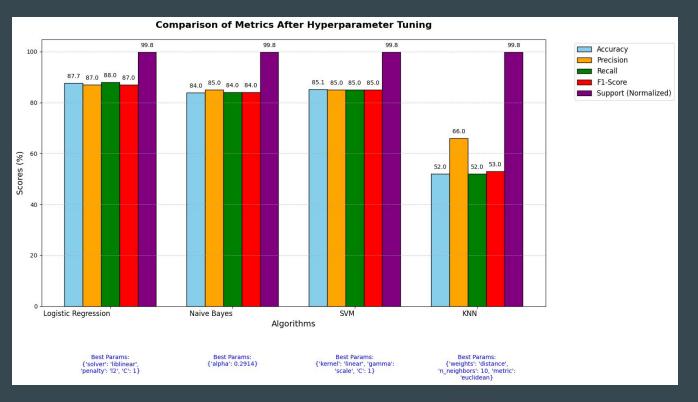
#### → XGBoost

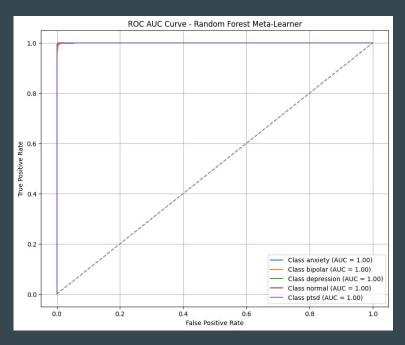
Delivered the highest accuracy and efficiency. The model demonstrated excellent handling of imbalanced datasets with robust predictions.

#### → LSTM

Outperformed traditional methods for sequential data. Captured contextual and temporal information effectively, achieving competitive results with complex text patterns.





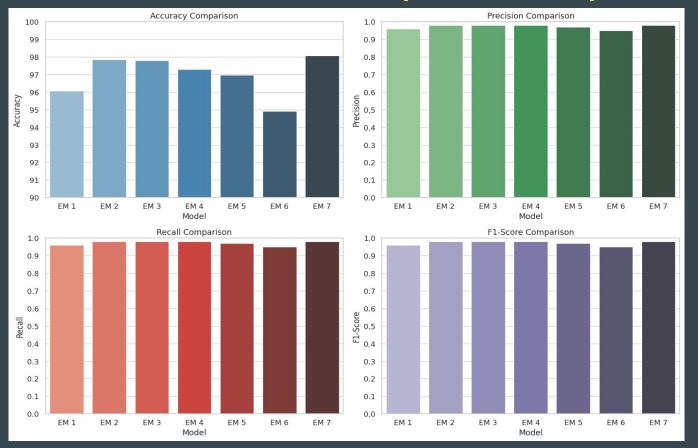


Confusion Matrix - Random Forest Meta-Learner 2000 3 anxiety - 1750 bipolar 361 19 1500 - 1250 True label 379 depression 1000 750 10 0 2126 normal 500 250 390 ptsd Predicted label

ROC AUC CURVE

CONFUSION MATRIX

ENSEMBLE MODEL (BASE MODELS : LOGISTIC REGRESSION, NAIVE BAYES, SVM, XGBOOST, LSTM, TRANSFORMER & META LEARNER : RANDOM FOREST) WITH ACCURACY OF 98.03% WAS USED IN WEB APPLICATION



COMPARISON OF
DIFFERENT ENSEMBLE
MODELS

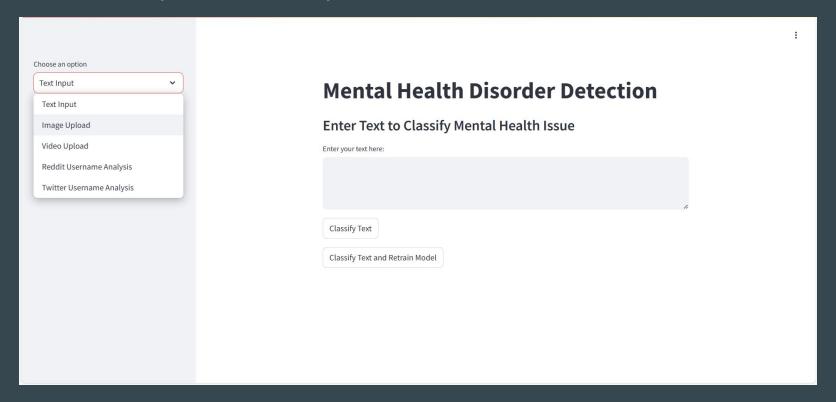
#### **PROTOTYPE**

- → A functional prototype was developed to classify mental health concerns from social media posts.
- → The system allows users to input text, upload images, or submit video for classification.
- Key functionalities include text preprocessing, feature extraction, and prediction using trained machine learning models.
- Real-time results display the most probable mental health concern with confidence scores.
- → Allow user to retrain model.

- → Text Classification : Users can directly input text for immediate analysis and classification.
- → Image-based Classification :

  Extracts text from uploaded images using OCR (pytesseract), get captions and facial emotions to classify the content.
- Video-based Classification: Processes uploaded video files, extract text from frames, transcribes speech to text, get captions and facial emotions to classify the content.
- Userprofile Analysis: Enables analysing user profiles based on posts in Reddit and Twitter

### PROTOTYPE (CONTINUED)



### CONCLUSION

- → Significance of the Project

  Developed a robust system for early detection of mental health disorders through social media analysis.
- → Effective Use of Machine Learning

  Leveraged various machine learning models, identifying SVM as the most accurate for sentiment classification.
- → Impact on Mental Health Awareness

  Provides valuable insights for mental health professionals and public health organizations, enabling proactive interventions.
- → Potential for Future Development

  Future enhancements with deep learning and multimodal data can lead to even better accuracy and insights.
- → Commitment to Ethical Practices

  Emphasizes the importance of user privacy and ethical considerations in handling sensitive mental health data.

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## THANK YOU