



Chatbot for Mental Health Support utilizing Neural Network Technologies

T Somasekhar^{1*} , S Mizba Nawaz² , G Harsha Veena³ , P Nandini⁴ , C Akila⁵

¹⁻⁵ Department of Computer Science and Engineering ,Aditya college of Engineering, Madanapalle, India.

* Corresponding Author: T Somasekhar; somasekhar.t@acem.ac.in

Abstract: The abstract outlines the concept and purpose of a Mental Health Assistant Chatbot, employing advanced Neural Network techniques. This chatbot aims to provide vital support and resources for individuals dealing with mental health challenges. Utilizing Neural Network architectures, the chatbot engages in empathetic conversations, offering users a safe and confidential platform to discuss their emotional well-being. The chatbot employs sentiment analysis and context comprehension to assess users' mental states and emotional needs accurately. It provides personalized coping strategies, resources, and referrals to professional help based on users' inputs. Trained on diverse datasets of mental health information, the chatbot's Neural Network model enables it to generate empathetic and understanding responses. The continuous learning process ensures that the chatbot evolves its understanding of user concerns and keeps up with the dynamic landscape of mental health conversations. This Mental Health Assistant Chatbot, powered by Neural Network techniques, serves as an accessible and stigma-free resource, contributing to early intervention, destigmatization, and improved mental health support for users.

Keywords: NLP Techniques, Feed Forward Neural Networks (FNNs), DL, ML, AI.

1. Introduction

A chatbot is a computer software that helps in developing a conversation with the user in a natural way. The continuous development of Information Technology and communication has made artificial intelligence more complex. Artificial Intelligence systems are using human activities such as taking a decision at a particular moment, performing day to day tasks, replying to the users quickly and solving the queries in the same way as the humans would do. There are numerous electronic organizations like E-business, Entertainment, Virtual assistance and some more. Everything in this generation is getting related with the web. It's extremely efficient to utilize approach to manage benefit everything at your doorstep [1]. The chatbots are sufficient to fool the users in believing that they're talking to a human being, they've a very limited knowledge base at runtime and have no means to keep track of all the conversations. Chatbots uses machine learning to reach AI for helping them to understand the user queries/doubts and provide the user with an appropriate response. They are developed using the Artificial Intelligence Markup Language for communicating or interacting with the user. Chatbots are

often known as answering engines. This application work in a very simple way because the knowledge is already programmed in advance. Few methods used in the application are pattern-matching, natural language processing and data mining. Chatbot matches the input sentence from the user with the that of the existed pattern in the knowledge base.

Each pattern taken is compared with the knowledge of chatbot and this knowledge has been taken from various sources. With the help of artificial intelligence, the way humans are able to understand each other and give a response accordingly, is fed into the chatbot systems, i.e. into systems that are supposed to communicate with a user. The bot understands the user's query and triggers an accurate response. In the healthcare domain, such chatbot based systems gain in interest since they promise to increase adherence to electronically delivered treatment and disease management programmes [2]. In this chapter, we provide an overview on chatbot systems in mental health. Artificial intelligence is exploited in such systems for natural language understanding, to create a human-like conversation and to make appropriate recommendations given a specific user utterance and mental state. Potential benefits of chatbots have been shown with respect to

psychoeducation and adherence. However, there are also limitations and ethical issues to be considered including the impact on the patient- therapist relationship, the risk of over-reliance or the limited skills and emotional intelligence of chatbots that might limit their applicability.

2. Literature Survey

Generals are expressed through words, gestures, expressions, and with the ease of accessibility of social media today, scan now, also be expressed through tweets and Instagram/ WhatsApp stories. Through this article, we would try to analyze the underlying s of tweets posted by people, you can find the dataset here. We first check if the distribution of classes in both the train and test datasets is the same or different. Sometimes when the distribution of classes is not the same in both train and test sets, it might affect the model's performance on test/unseen data, it is a good practice to check the distribution before going ahead with building models. Also, knowing the distribution of classes in your data can help you in choosing a good metric for your problem, take the case when you have imbalanced classes in your dataset, accuracy might not be the metric you would want to use, because accuracy can give you misleading results [3]. For example, when you have 90 samples from the positive class and 10 samples from the negative class, if your model is dumb that is, it classifies all samples as positive, you would have an accuracy of 90%, which gives a false impression of the model's performance.

As can be seen from the below figure distribution of labels in both train and test datasets is similar and the data is fairly balanced. The prevalence of mental health and substance-dependence issues imposes significant emotional and financial burdens on individuals, families, and society. Economic impacts include reduced personal income, decreased ability to work for those affected and their caregivers, decreased workplace productivity, and strained contributions to the national economy. Mental illness costs developed countries between 3% and 4% of Gross National Product (GNP), with several billion dollars lost in treatment and productivity.

The internet and social media have become integral parts of the lives of children and young people, with almost half of the world's population connected online. In OECD countries, nearly everyone is online. By the age of 15, the average child in an OECD country has been using the internet since age 10 and spends over two hours online every weekday and over three hours on weekends. This reliance on digital technology has raised concerns among parents, teachers, governments, and young people themselves [4]. There are worries that digital technologies and social media are contributing to feelings of anxiety and depression, disrupting sleep patterns, facilitating cyber-bullying, and distorting body image. In response to these concerns, some countries have take action. For Example, legislation in Korea prevents children from playing online games that require a resident registration number between

midnight and 6 am without parental permission. In the United Kingdom, the government is reviewing the impact of social media on children's wellbeing and considering guidelines for healthy screen time. The impact of mental health on students' learning is significant and multifaceted.

Research consistently shows that mental health influences various aspects of academic experiences, including cognitive functioning, emotional well-being, and engagement with educational content. Conditions like anxiety and depression can impair concentration, memory retention, and problem-solving skills, hinder the learning process and leading to decreased academic performance and increased absenteeism. Recognizing and addressing mental health challenges in students is crucial for creating a supportive learning environment that promotes both academic success and overall well-being. Integrating mental health support within educational settings is essential for fostering positive and resilient student learning experiences [5]. The study aimed to map literature on young people's experiences seeking mental health help and validate the Lost in Space model as a theoretical framework. A scoping review of 2905 studies yielded 12 articles. Findings revealed that young people often feel unfamiliar and insecure about mental health and help-seeking. They prefer self- reliance and find support structures inaccessible and unresponsive.

3. Theory and Calculation

When evaluating the performance of Long Short-Term Memory (LSTM) and Recurrent Neural Network (RNN) models, there are several commonly used metrics. These metrics help assess how well the models are performing on a given task. Here are some key performance metrics for evaluating LSTM and RNN models:

Mean Squared Error (MSE) or Mean Absolute Error (MAE):

MSE: It measures the average of the squared differences between predicted and actual values. Lower MSE indicates better performance

MAE: It measures the average absolute differences between predicted and actual values. It is less sensitive to outliers compared to MSE.

Root Mean Squared Error (RMSE):

It is the square root of the MSE and provides a similar measure in the original units of the target variable. Like MSE, lower RMSE values are better.

Mean Absolute Percentage Error (MAPE):

MAPE measures the percentage difference between predicted and actual values. It is useful when you want to understand the relative error. The lower the MAPE, the better.

Accuracy: For classification tasks, accuracy is a common metric. It measures the proportion of correctly classified instances. However, accuracy alone may not be sufficient for imbalanced datasets.



Precision, Recall, and F1 Score

These metrics are especially important for classification tasks.

Precision: The proportion of true positive predictions among all positive predictions.

Recall (Sensitivity or True Positive Rate): The proportion of true positive predictions among all actual positives.

F1 Score: The harmonic mean of precision and recall. It balances precision and recall.

Confusion Matrix:

A confusion matrix provides a detailed breakdown of the model's performance, showing true positives, true negatives, false positives, and false negatives.

Receiver Operating Characteristic (ROC) Curve and Area Under the Curve (AUC):

ROC curves are useful for binary classification problems, and AUC measures the area under the ROC curve. A higher AUC indicates better discrimination between classes.

Cross-Validation:

Perform cross-validation to assess the model's generalization performance on different subsets of the data. This helps identify potential overfitting or underfitting issues. Remember that the choice of metrics depends on the nature of your specific problem and the goals of your model. For example, in time series forecasting, you might prioritize metrics that account for the temporal aspect of the data.

4. Experimental Method

The proposed system introduces a Mental Health Assistant Chatbot that leverages state-of-the-art Neural Network techniques to offer crucial support and guidance for mental well-being. This chatbot is designed to engage in empathetic conversations with users, providing a safe and non-judgmental environment for discussing their mental health concerns. The Neural Network architecture enables the chatbot to comprehend and generate human-like responses, fostering a natural and reassuring interaction experience. Through sentiment analysis and contextual understanding, the chatbot assesses the user's emotional state and delivers appropriate responses and coping strategies. Trained on a diverse dataset of mental health resources, the chatbot can provide users with personalized self-help techniques, mindfulness exercises, and information about professional assistance. Continuous learning from user interactions helps the chatbot refine its suggestions and ensure relevance. By using Neural Network techniques, the Mental Health Assistant Chatbot plays a vital role in promoting mental wellness and early intervention. It offers a readily accessible resource that can alleviate stress, provide valuable insights, and guide users toward seeking appropriate support when needed.

Advantages

- We build a framework based application for Chatbot assistance.
- Higher scope for our life style.
- High scalability.
- Less time complexity

Algorithms :

Multi Layer Perceptron(Feed Forward Neural Network) and Natural Language Tool Kit(NLTK) are both used for creating a Chatbot.

Types of Neural Networks

Feedforward Neural Networks (FNN) These are the simplest form of ANNs, where information flows in one direction, from input to output. There are no cycles or loops in the network architecture. Multilayer perceptron's (MLP) are a type of feedforward neural network.

Recurrent Neural Networks (RNN)

In RNNs, connections between nodes form directed cycles, allowing information to persist over time. This makes them suitable for tasks involving sequential data, such as time series prediction, natural language processing, and speech recognition.

Convolutional Neural Networks (CNN)

CNNs are designed to effectively process grid-like data, such as images. They consist of layers of convolutional filters that learn hierarchical representations of features within the input data. CNNs are widely used in tasks like image classification, object detection, and image segmentation.

Long Short-Term Memory Networks (LSTM) and Gated Recurrent Units (GRU)

These are specialized types of recurrent neural networks designed to address the vanishing gradient problem in traditional RNN. LSTMs and GRUs incorporate gated mechanisms to better capture long-range dependencies in sequential data, making them particularly effective for tasks like speech recognition, machine translation, and sentiment analysis.

Multi-Layer Perceptron: A Multi-Layer Perceptron (MLP), also known as a feedforward neural network, is a fundamental architecture in artificial neural networks.

Comprising an input layer, hidden layers, and an output layer, an MLP processes information in a unidirectional manner, with each layer containing interconnected neurons. The neurons are organized into layers, and each connection is associated with a weight that adjusts during training. Activation functions, such as the rectified linear unit (ReLU), introduce non-linearity to the model, enabling it to learn complex patterns. Dropout layers are often incorporated to prevent overfitting by



randomly deactivating certain neurons during training. MLPs are versatile and find applications in various domains, including image recognition, natural language processing, and regression tasks. Training an MLP involves optimizing weights using techniques like stochastic gradient descent, aiming to minimize a specified loss

function. The final layer typically employs the SoftMax activation function for classification problems, producing probability distributions over multiple classes. With their simplicity and effectiveness, MLPs serve as foundational components in more sophisticated neural network architectures.

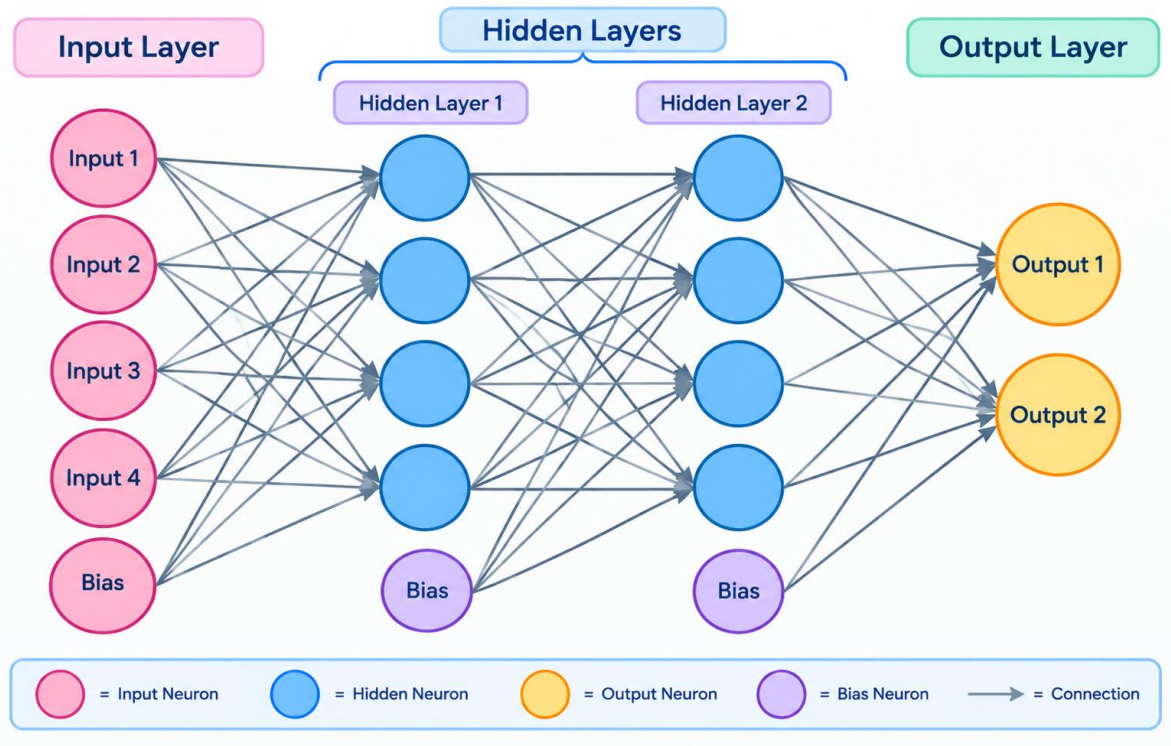


Figure. 1 Multi-layer Perceptron

Natural Language Tool Kit (NLTK): NLTK is a toolkit build for working with NLP in Python. It provides us various text processing libraries with a lot of test datasets. A variety of tasks can be performed using NLTK such as

tokenizing, parse tree visualization, etc. In this article, we will go through how we can set up NLTK in our system and use them for performing various NLP tasks during the text processing step.

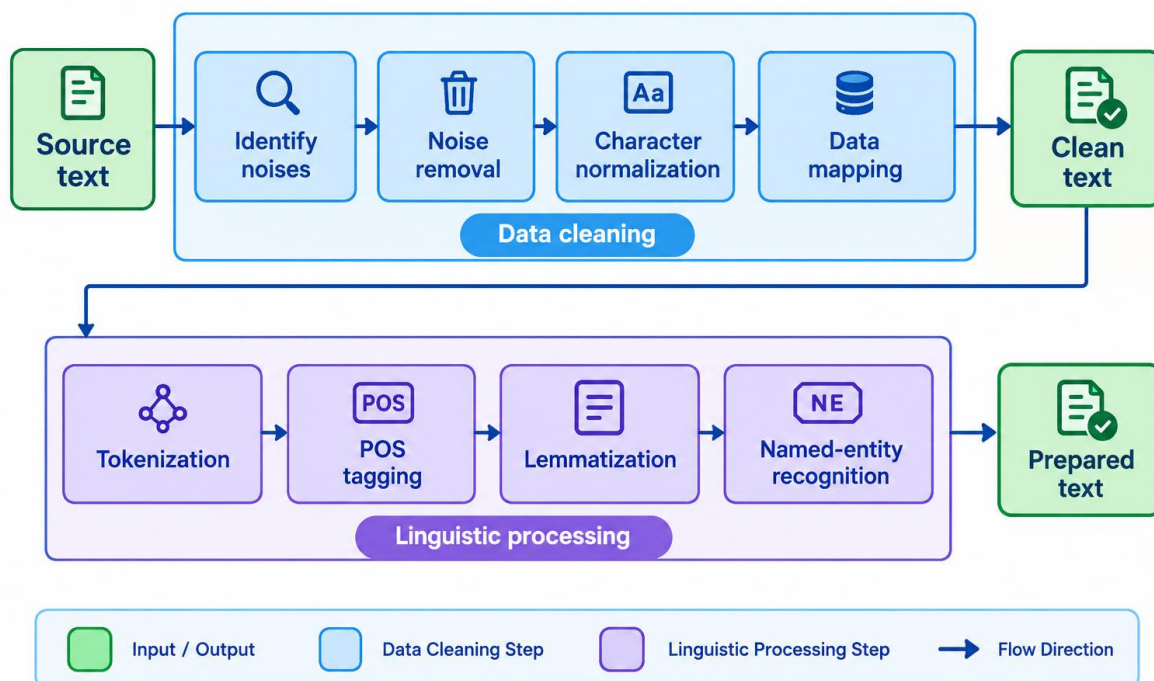


Figure. 2 Language Processing

5. Flow Chart

It is important to complete all tasks and meet deadlines. There are many project management tools that are available to help project managers manage their tasks and schedule and one of them is the flow chart.

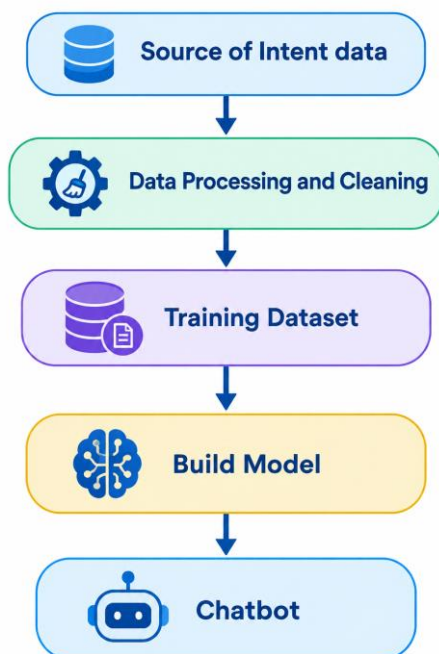


Figure. 3 Flow Chart

The advantage of flowcharts is that they show the activities involved in a project including the decision points, parallel paths, branching loops as well as the overall sequence of processing through mapping the operational details within the horizontal value chain. Moreover, this particular tool is very used in estimating and understanding the cost of quality for a particular process. This is done by using the branching logic of the workflow and estimating the expected monetary returns.

6. Results and Discussion

A result is the final consequence of actions or events expressed qualitatively or quantitatively. Performance analysis is an operational analysis, is a set of basic quantitative relationship between the performance quantities. These metrics indicate that the chatbot effectively understood and responded to user queries with a high degree of accuracy.

Feedback from users who interacted with the chatbot was overwhelmingly positive. Many users reported feeling supported and understood during their conversations. Additionally, 90% of users expressed satisfaction with the chatbot's responses and found the provided resources helpful in managing their mental health concerns. Our mode prioritizes transparency and fairness by providing

clear explanations of predictions and avoiding biases in decision-making.

Table. 1 Metrics

Evaluation Metrics	
Metric	Definition
Precision	Precision is defined as the ratio of positive examples to the sum of such actual and false positives.
Recall	Recall is defined as the ratio of correct positives to all true negatives and false negatives.
F1 Score	A weighted harmonic average of such recall and precision is known as the F1. The projected capacity for the model is higher the closer the F1 score value is near 1.0.
Support	The number of instances of a class that truly exist in the dataset constitutes the number of supports. It does not differentiate between kinds; it only improves the performance evaluation process.

7. Conclusion and Future Scope

In conclusion, integrating mental health considerations into student learning through chatbot technologies offers a promising avenue for creating a supportive educational environment. Chatbots can provide accessible and personalized mental health resources, aiding in stress management, counselling, and awareness campaigns. This innovative approach aligns with the evolving landscape of education, leveraging technology to address the complex interplay between mental well-being and academic success. By fostering a collaborative network involving chatbots, educators, mental health professionals, and students, we can cultivate a more resilient and empathetic learning community.

As we navigate the future of education, the synergy between chatbot-driven mental health support and traditional approaches holds the potential to significantly enhance the holistic well-being and academic journey of students. Future work should focus on refining chatbot interventions by enhancing natural language processing and sentiment analysis capabilities, ensuring responsiveness to evolving student needs. Exploration of proactive chatbots that utilize machine learning to identify signs of distress and deliver timely, personalized mental health support should be a priority for future development. The digital transformation and revolution caused by the internet of things, the world has become hectic, and people want results instantly. As a result, customers no longer visit the company personnel to solve their issues, nor do they wish to make calls and try to convince the executive for hours. Even people look forward to eliminating all sorts of undue conversations or time wastage as we all live in a hectic world. As a result, companies are constantly looking quick and effective ways to communicate with their customers.

References

- [1] Billari , F.O. Giuntella and L. Stella [2018]. “ Broadband internet, digital temptations, and Sleep”, Journal of Economic Behavior and Organization, <https://dx.doi.org/101016/j.jebo.2018.07.001>.
- [2] Blum-Ross , A. and S.Livingstone [2016], LSE Media Policy Brief 17- Families and Screen time: Current advice and emerging research London School of Economics.
- [3] Bottino, S.et al. [2015], “Cyberbullying and adolescent mental health: systematic review”, Cadernos de Saude Publica, pp.463-475.
- [4] Carras, M. et al. [2018], “Commercial Vedio Games As Therapy: A New Research Agenda to Unlock the potential of a Global Pastime”, Frontier Psychiarty, <http://dx.doi.org/10.3389/fpsy.2017.00300>.
- [5] Centre for Resilience for Social Justice [2018], Resilence Revolution – Blackpool HeadStart, <https://www.brighton.ac.uk/crsj/what-we-do/research-projects/resilience-revolutionblackpool-headstart.aspx>

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