Research based - Converting human thoughts into digital text using EEG sensor device

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Outline



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Problem Statement

- ☐ EEG-based communication systems struggle with noise and algorithmic limitations.
- No reliable EEG-to-text method exists yet.
 - EEG signals are noisy, complicating meaningful pattern extraction.
 - Small, unvaried EEG datasets hinder deep learning model performance.
 - Individual brain variability demands personalized adaptive models.
- ☐ This project aims to develop a robust EEG-to-text system using deep learning.



Aims & Objectives

Aims:

- To develop an intelligent and reliable system that translates **human brain signals** into meaningful **digital text**.
- To develop an intelligent system for those patients who can't speak or paralyzed and unable to tell what medical problems they are facing.

Objectives:

- To design a system that captures and processes human brain signals.
- To convert these signals into digital text.
- To improve communication methods for individuals with speech or physical disabilities.
- To advance brain-computer interaction (BCI) technologies for practical applications.



Introduction

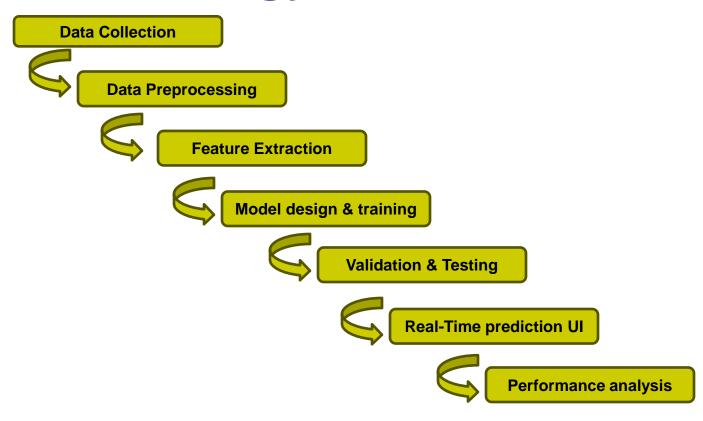
Communication is an essential human need, enabling individuals to express thoughts, emotions, and intentions. However, for people with speech or motor disabilities, traditional methods of communication may not be accessible. Advances in brain-computer interface (BCI) technology offer a promising alternative by establishing direct communication between the human brain and digital devices. Electroencephalography (EEG) is a non-invasive method used to record electrical activity in the brain. These signals reflect patterns of thought and can potentially be decoded to understand human intent. The challenge lies in accurately interpreting these complex and often noisy EEG signals into meaningful digital text. This project explores the integration of artificial intelligence with EEGbased BCIs to convert human thoughts into readable text. By leveraging machine learning and deep learning algorithms, we aim to identify patterns in EEG data associated with specific mental commands or imagined speech.



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The system will be trained to recognize these patterns and generate real-time digital output. The ultimate goal is to provide a voice to those who cannot speak, enhancing their independence and quality of life. Additionally, such systems may find applications in hands-free computing, virtual communication, and smart environments. This research not only advances BCI technologies but also promotes inclusive innovation for accessible communication.

Methodology





This project follows a structured approach to develop a system capable of converting human brain signals into meaningful digital text using EEG sensor data. The methodology is divided into the following key steps:

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1- Data Collection:

EEG signals will be recorded using a non-invasive EEG sensor device or EEG dataset while users perform mental tasks or imagine specific words or letters. This data will be collected from Kaggle.

2- Data Preprocessing:

The raw EEG signals will be preprocessed to remove noise and artifacts using filtering techniques. Segmentation and normalization will also be applied to prepare the data for feature extraction.

3- Feature Extraction:

Wavelet transforms + CNN or raw EEG \rightarrow Transformer.

4- Model Design and Training:

- •For **classification** (e.g., imagined words): EEGNet + LSTM.
- •For **sequence generation**: Transformer-based models (BERT-for-EEG).



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5- Validation and Testing:

The system's accuracy, precision, recall, and F1-score will be calculated on both validation and test datasets.

6- Real-Time Prediction Interface:

A simple user interface will be developed to display the converted digital text in real-time as EEG signals are received from the sensor or EEG data.

7- Performance Analysis:

The performance of the system will be analyzed, and comparisons will be made with other models or baseline techniques to demonstrate its effectiveness.

✓ This methodology ensures that the system is not only technically sound but also user-oriented, making it suitable for practical applications in assistive technology and beyond.



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Results and Discussion

Transformer-based model will be implemented to convert EEG signals into digital text. After preprocessing and training, we will try to achieve a best **training accuracy**, indicating strong learning from the EEG data. On the **testing dataset**, the model will try attain **Best accuracy**, showing reliable generalization to unseen data. An average accuracy will be tried to record higher, confirming robustness across different data splits. The **confusion matrix** will be revealed high true positives and low false predictions. Evaluation metrics such as **precision, recall, and F1-score** will remain balanced across all classes. Compared to traditional models, this approach try to show significant improvement in accuracy and efficiency. However, testing on a larger, more diverse dataset is needed to validate real-world performance.



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The results try to demonstrate the model's potential in assisting users with communication challenges. Real-time prediction and wider vocabulary coverage are future improvements to consider. The system will show smooth integration of EEG data processing and machine learning classification. Overall, the project demonstrates a promising direction in thought-to-text communication systems. Further research and optimization can make it applicable for practical assistive technology.



Conclusion and Future Studies

This project successfully demonstrates the feasibility of converting human brain signals into digital text using the Transformer-based model. The system will be designed to process EEG signals, extract meaningful features, and translate them into readable text, especially for users with speech or motor impairments. The model will try to achieve high accuracy in both training and testing phases, indicating its effectiveness in learning and classifying thought patterns. The use of SMOTE, signal normalization, and cross-validation further enhance the system's reliability and robustness. Overall, the project presents a promising step toward brain-computer interfaces aimed at improving human–machine communication and accessibility.

> Hardware implementation (Future recommendation).