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## **ABSTRACT**

Cyberbullying Detection uses a combination of machine learning techniques such as TF-IDF vectorization, logistic regression, multilayer perceptron, CNNs and LSTM networks to create a robust model for detecting cyberbullying. By Employing BERT model, it's able achieve higher accuracy and better performance in identifying offensive content on social media platforms.

The existing System for detecting cyberbullying in Indian Language Bengali on social media. The model uses text preprocessing, TF-IDF, and Instance Hardness Threshold (IHT) for resampling, it uses multiple Machine learning algorithms for detection of online harassment.

However, the existing System does not address the practical challenges like Real-Time Detection and the Technique used for Resampling deduce the actual size of dataset to balance the dataset which leads to lower accuracy rate. To overcome these Limitations, the proposed system uses the BERT model, known for its advanced contextual understanding and bidirectional processing capabilities, to enhance prediction accuracy.

### **KEYWORDS:**

Cyberbully, Machine learning, Social media, NLP, Twitter, Facebook.

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## LIST OF ABBREVIATIONS

1. **AI** - Artificial Intelligence
2. **ANN** - Artificial Neural Network
3. **BERT** - Bidirectional Encoder Representations from Transformers
4. **Bi-LSTM** - Bidirectional Long Short-Term Memory
5. **CNN** - Convolutional Neural Network
6. **CML** - Conventional Machine Learning
7. **DNN** - Deep Neural Network
8. **GPU** - Graphics Processing Unit
9. **IDE** - Integrated Development Environment
10. **IHT** - Instance Hardness Threshold
11. **LIWC** - Linguistic Inquiry and Word Count
12. **ML** - Machine Learning
13. **MLP** - Multi-Layer Perceptron
14. **NLP** - Natural Language Processing
15. **RNN** - Recurrent Neural Network
16. **RoBERTa** - A Robustly Optimized BERT Pretraining Approach
17. **SVM** - Support Vector Machine
18. **TF-IDF** - Term Frequency-Inverse Document Frequency
19. **VCS** - Version Control Systems

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