SR. No	INFORMATION ABOUT THE PAPER	LEARNING OUTCOME
1.	Title: Automatic Detection of Cyberbullying in Social Media Text.	Data-Collection: Two corpora were constructed by collecting data from social networking sites like Ask.fm.
	Publication Name: PLOS ONE Journal Date of Publish: 2018 Authors: Cynthia Van Hee, Gilles Jacobs, Chris Emmery, Bart Desmet, Els Lefever, Ben Verhoeven, Guy De Pauw, Walter Daelemans, and Veronique Hoste Motive: Their focus was on automatic cyberbullying detection in social media text by modelling posts written by bul lies, victims by standards of online bullying.	Pre-processing: Tokenization, PoS-tagging, and lemmatization were used Word n-grams and lexicon features were extracted. Model: Models developed for English and Dutch to test for language conversion and subsequent accuracy. Classified using SVM which gave the accuracy for the English language - 64% and for the Dutch language - 61%.
2.	Title: Predicting Cyberbullying on Social Media in the Big Data Era Using Machine Learning Algorithms. Publication Name: IEEE Access Date of Publish: 2019 Authors: MOHAMMED ALI AL-GARADI, NAWSHER KHAN, GHULAM MURTAZA, IHSAN ALI, HASAN ALI KHATTAK, AND ABDULLAH GANI Motive: Their proposed was to reduce textual cyberbullying because it has become the dominant aggressive behaviour in social media sites.	 Data Collection: Extracted data from Wikipedia, you-tube, Twitter and more social media site. Pre - processing: They applied Tokenization, PoS-tagging, and lemmatization & N-gram was used up to 5 levels to calculate TFIDF and count vector. Model: Classified using > SVM, K clustering, Random forest, Decision Trees, Naïve Bayes.
3.	Title: Detecting Cyberbullying and Aggression in Social Commentary using NLP and Machine Learning. Publication Name: International Journal of Engineering Technology Science and Research (IJETSR) Date of Publish: 2018	 Data Collection: Data set obtained from Wikipedia, YouTube, Twitter. Pre – processing: Count Vectors and TF-IDF vectors created by defining n-gram of up to 5 levels. Model: Logistic Regression, SVM, Random Forest, and Gradient Boosting.

	Authors: Kshitiz Sahay, Harsimran Singh Khaira, Prince Kukreja, Nishchay Shukla Motive: They identified and classified bullying in the text by analyzing and studying the properties of bullies and aggressors and what features distinguish them from regular users.	
4.	Title: Unsupervised Cyber Bullying Detection in Social Networks Publication Name: IEEE Date of Publish: December 4-8, 2016 Authors: Michele Di Capua, Emanuel Di Nardo, Alfredo Petrosino Motive: Was to design a model inspired by Growing Hierarchical SOMs, which is able to efficiently cluster documents containing bully traces, built upon semantic and syntactic features of textual sentences.	Pre – Processing: Followed an Unsupervised approach with Syntactic, Semantic, Sentiment, and Social features extracted. stop word removal, punctuation removal done to generate word clusters. Model: CNN was applied using the Kohonen map (or GHSOM). To measure the goodness of generated data clusters, the F1 Score is measured.
5.	Title: Detection of Cyberbullying Incidents on the Instagram Social Network Publication Name: Arxiv Date of Publish: 2015 Authors: Homa Hosseinmardi, Sabrina Arredondo Mattson, Rahat Ibn Rafiq, Richard Han, Qin Lv, Shivakant Mishra Motive: Automatic Cyberbullying Detection in Instagram text by modelling post written by bullies.	 Deciding posts based on shortlisting words of a caption. Using IP on Instagram posts for deciding emotional response or test response in case of text pictures.
6.	Title: Detection of Cyberbullying Using Deep Neural Network Publication Name: IEEE Date of publish: 2019	Data-Collection: twitter dataset which consists of 69874 tweets. Pre-processing: stop words, accentuation marks and lowercasing.

	Authors: Vijay Banerjee, Jui Telavane, Pooja Gaikwad, Pallavi Vartak Motive: Their focus was to develop the model for cyber bullying detection using deep neural networks.	Model: CNN. The accuracy of this model was 93.97% which is greater than the other models.
7.	Title: Cyberbullying Classification using Text Mining. Publication Name: IEEE Date of publish: 2017 Authors: Noviantho, Sani Muhamad Isa, Livia Ashianti Motive: Their focus was to create a classification model for cyberbullying using Naive Bayes method and Support Vector Machine (SVM).	Data Collection: Collected from Kaggle which provides 1600 conversations in Formspring.me. Pre - processing: They applied tokenization, transfer case, stop word removal, filter token, stemming, and generating n-gram. Model: Naive Bayes method and Support Vector Machine (SVM).
8.	Title: Hate Speech on Twitter: A Pragmatic Approach to Collect Hateful and Offensive Expressions and Perform Hate Speech Detection. Publication Name: IEEE Date of publish: 2018 Authors: Hajime Watanabe, Mondher Bouazizi and Tomoki Ohtsuki Motive: Their focus was to detect hate speech on Twitter using of technique is based on unigram and pattern that automatically trained from the dataset.	Data Collection: They use 3 types of datasets: 1) the dataset from crowd flower contains 14000 tweets are classified into clean, offensive and hateful. 2)also from crowd flower tweets classified into offensive, hateful and neither 3) third dataset was from GitHub in which tweets were classified into sexism, racism and neither. Pre – processing: removed URL and tags from tweets also they did tokenization, Part of Speech Tagging, and lemmatization Model: binary classification and ternary classification to identify sentiment-based features, semantic features, Unigram features and pattern feature
9.	Title: Cyberbullying Detection using Pre-Trained BERT Model Publication Name: IEEE	Dataset: Form spring (a Q&A forum) and Wikipedia talk pages (collaborative knowledge repository). Form spring dataset contains 12773 question-answer pair comments of

Date of publish: 2020

Authors: Jaideep Yadav, Devesh

Kumar, Devesh Kumar

Motive: To develop a model to classify

cyberbullying using pre-trained BERT model

which 776 are bully posts. Wikipedia dataset contains 115864 discussion comments which are manually annotated by ten persons of which 13590 comments which are labelled as bully.

Pre-processing: The dataset was pre-

processed.

Model: pre-trained BERT model. The model gave accuracy of 94% for formspring dataset which is oversampled 3 times and 81% for wikipedia dataset.

10. **Title:** Detecting Hate Speech and Offensive Language on Twitter using Machine Learning.

Publication Name: IJCMS

Date of Publish: April- 2021

Authors: S.E.VISWAPRIYA, AJAY GOUR,

BOLLOJU GOPI CHAND

Motive: Detecting tweets as hateful, offensive and

clear.

Data: Crowd Flower and GitHub. Tweets were fetched by the tweet id using twitter API. These datasets were then combined.

Pre-processing: Tweets were converted to lower case and Space Pattern, URLs, Twitter Mentions, Retweet Symbols and Stop words were removed. To reduce inflectional forms of words, stemming was applied. The dataset was then spited into 70% training and 30% test samples.

Feature Extraction: n-gram features from the tweets were extracted and were weighed according to their TFIDF values. Unigram, Bigram and Trigram features along with L1 and L2 normalization of TFIDF were considered.

Models: Logistic Regression, Naïve Bayes and Support vector machine were used. 95% accuracy was obtained using Logistic Regression with L2 Normalization and n=3.

11. **Title:** Detection of Hate Tweets using Machine Learning and Deep Learning.

Publication Name: IEEE

Date of Publish: February 2021

Authors: Lida Ketsbaia, Biju Issac and Xiaomin

Chen

Motive: To detect hateful and offensive tweets.

Data: A dataset created by the University of Maryland and Conrell University of about 35000 and 24000 tweets respectively was used with tweets labelled as Hate and Non hate.

Pre-processing: tweets were converted into lowercase, numbers, URL's and user mentions, punctuations, special characters and stop words were removed and contradictions were replaced. Dataset was then balanced.

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