



PREDICTING SOCIAL MEDIA CYBERBULLYING WITH MACHINE LEARNING ALGORITHMS IN THE AGE OF BIG DATA

Rakesh Pandey, Assistant Professor, Department of Electronic & Communication Engineering, Birla Institute of applied sciences, Bhimtal

Neeraj Kumar Bahuguna, Assistant Professor, Department of Electrical Engineering, Nanhi Pari Seemant Engineering Institute, Pithoragarh

Harendra Singh Bohara, Assistant Professor, Department of Applied Sciences, Birla Institute of applied sciences, Bhimtal

Manish Bhatt, Assistant Professor, Department of Computer Science Engineering, Birla Institute of applied sciences, Bhimtal

ABSTRACT

The widespread usage of social media platforms in this age of big data has made it easier for people from different cultures to have conversations with one another and broadened human contact beyond the confines of conventional limits. However, along with these developments, there has been an alarming surge in online hatred and cyberbullying, which poses enormous obstacles for the purpose of ensuring that the online environment is both secure and welcoming to all users. This research study proposes an innovative method for anticipating and fighting hostile behaviour in social media by making use of machine learning algorithms. The method was developed by the authors of the research paper. The article includes many case studies that might serve as models for application across a variety of social media platforms, as well as emphasises the need of establishing prediction models to deal with aggressive behaviour. It looks into the one-of-a-kind issues that come along with the process of creating cyberbullying prediction models that are geared particularly to social media platforms. The research not only gives insights into the process of recognising cyberbullying, but it also offers a thorough framework for successfully combating it. These benefits are both provided by the study. The selection of algorithmic features and the implementation of a variety of machine learning algorithms to predict cyberbullying behaviours have been the focal points of attention in recent years, notwithstanding the progress that has been made in data collecting and feature engineering. The results of this study have revealed previously unknown challenges and obstacles, which opens up fruitful new lines of inquiry and warrants future examination. This study highlights the relevance of employing machine learning algorithms and big data analysis to foster a safer and more inclusive online environment by adding to the increasing body of knowledge in the area of social media cyberbullying prediction. This work adds to the expanding body of knowledge in the field of social media cyberbullying prediction.

Keywords: Social media, cyberbullying, machine learning algorithms, big data, prediction models, online hostility, algorithmic feature selection, safe online environment.

I. Introduction

In this day and age of big data and information and communication technology (ICT), the dynamics of social connections have been altered, which has resulted in the dismantling of cultural barriers and made it simpler for individuals from diverse backgrounds to conduct conversations with one another [1]. In contrast to earlier eras, when it was difficult to communicate due to time and geographical limitations, the current breakthroughs in technology have fundamentally revolutionised the process of sharing ideas and information on a global scale [1]. This is in contrast to earlier eras, when it was difficult to share information and ideas because it was difficult to share information because it was difficult to share ideas. In particular, the widespread usage of social media platforms, which are commonly referred to as SM (social media), has led in the generation of enormous volumes of user-generated content in addition to rich data documenting human activity and online networks [2].

On the other side, in spite of the many benefits brought about by social technologies, there has been a



growing concern over the rise of online hostility and cyberbullying [3]. Cyberbullying is defined as the act of harassing, threatening, or inflicting bodily harm on another person via the use of an electronic device [3, 4]. Examples of such devices include mobile phones, laptops, and tablets. It encompasses a broad spectrum of aggressive activities, including verbal abuse, defamation, sexual harassment, blackmail, insults, and being excluded from social occasions [3]. When it comes to the process of establishing a secure and inviting atmosphere online, the pervasive nature of cyberbullying provides a significant challenge that must be overcome.

In order to create prediction models that are capable of combating and preventing aggressive activity on social media platforms, researchers have resorted to using machine learning algorithms and big data analytics [4]. This is due to the fact that cyberbullying is a severe issue that has to be addressed as soon as possible. [4] Algorithms that use machine learning may consistently predict and detect patterns linked with cyberbullying and other undesirable human activities. These algorithms are able to do this by using the vast amounts of data that are available on social media sites. This allows them to more accurately predict user behaviour. This method makes it feasible to detect potential instances of cyberbullying before they really take place, which in turn makes it possible to take actions to prevent them from occurring.

Despite this, the creation of accurate cyberbullying prediction algorithms for social media sites has its own unique set of challenges [5]. Because of the large amount of data and the need to use computational approaches for feature selection, careful consideration and inventive problem-solving are required [5]. The purpose of this research study is to provide helpful insights into the process of constructing predictive models for the detection of cyberbullying in social media, with a special focus on the challenges that are specific to social media platforms [5].

The purpose of this study is to shed light on important aspects of the process of developing accurate and useful models for predicting instances of cyberbullying [5]. An in-depth research of data collection, feature engineering, and algorithmic selection will be used to achieve this goal. In addition, the results of this research contribute to a more complete understanding of the identification of cyberbullying and provide a comprehensive framework for effectively combating cyberbullying within the context of social media [5].

By employing recent advancements in machine learning algorithms and large data analysis, this initiative aims to increase our ability to detect, treat, and prevent cases of cyberbullying. The ultimate objective is to provide a space online that is friendlier, more welcoming, and less dangerous for all users. The findings of this study not only contribute to the academic understanding of cyberbullying prediction, but they also provide useful insights and recommendations for stakeholders who want to combat this pervasive issue on social media platforms.

II. PREVIOUS WORK

The results of previous studies on the subject of predicting and detecting cyberbullying have opened the way for a better knowledge of the intricacies and difficulties connected with this problem. Several research have investigated the possibility of developing accurate prediction models via the use of machine learning algorithms and big data analytics.

The use of machine learning algorithms to the problem of identifying instances of cyberbullying on social media platforms has been shown to have promise in previous research. In order to extract essential characteristics from textual data and predict instances of cyberbullying, researchers have used a variety of methods such as support vector machines (SVM), random forests, and deep learning models [1]. The findings of these models have been encouraging in terms of their ability to properly detect instances of cyberbullying based on language patterns, sentiment analysis, and contextual clues [2].

In addition, academics have looked at the possibility of incorporating methods of social network analysis into cyberbullying detection programmes. These studies have emphasised the need of studying network features, such as centrality measures and community detection, in order to get a better



understanding of the dynamics of cyberbullying [3]. This was accomplished by analysing the social connections and interactions among users. The incorporation of network information in machine learning models has been found to increase the ability to predict instances of cyberbullying and to identify prominent members within social networks [4].

In addition, academics have investigated the possible use of natural language processing (NLP) methods in the identification of instances of cyberbullying. Researchers have conducted research with the goal of detecting certain linguistic patterns, abusive language, and hostile speech that are indicative of cyberbullying behaviours [5]. These studies make use of NLP algorithms. These methods have showed the power of automatically classifying instances of cyberbullying and differentiating them from ordinary online conversation, which contributes to the creation of reliable prediction models.

Researchers have investigated, in addition to computational techniques, the role that user-generated material, metadata, and contextual information play in the identification of cyberbullying. Previous research has investigated the use of user characteristics including posting frequency, time of activity, and interaction patterns in order to construct predictive models [6]. Researchers have attempted to improve the accuracy and usefulness of cyberbullying detection algorithms by taking the aforementioned contextual elements into consideration. This allows for proactive intervention and prevention techniques to be developed.

There has been substantial progress made in this area, but there are still many obstacles to overcome in order to construct cyberbullying prediction models that are accurate and dependable. Researchers face continual hurdles due to the dynamic nature of online platforms, the ever-changing patterns of cyberbullying behaviours, and the complications connected with contextual interpretation [7]. Future research efforts need to address these problems by investigating novel methods, combining real-time data, and taking into consideration the many facets that comprise cyberbullying.

In general, past research has laid the groundwork for understanding cyberbullying and developing effective strategies to prevent it via the use of machine learning algorithms and big data analytics. Researchers hope that by harnessing the power of these technologies, they will be able to build online settings that are safer and more welcoming for all users. This will be accomplished by properly predicting and preventing incidents of cyberbullying. Additional developments in this area have the potential to improve upon previously established models, unearth previously unknown insights, and ultimately have a positive impact on the health of online communities.

III. PROPOSED WORK & RESULTS

Java Server Pages, more often referred to as JSP, is a technology that is a part of the Java standard set that allows you to build Java web application pages that are dynamic and data-driven. JSP is the most common abbreviation for Java Server Pages. Java Server Pages is another term that is often abbreviated as JSP. Java Server Pages, or JSP for short, is a platform that was developed on top of the Java Servlet standard. The vast majority of the time, the two technologies are compatible with one another; this is notably the case with Java web applications that are much older than the generation that is now in use. When you are working with servlets, you will begin by writing code in Java, and after that is complete, you will embed client-side content (like HTML) inside that code. When dealing with JSP, on the other hand, you first need to create the script or markup that will be shown on the client's side of the application. After that, you need to include JSP tags into your page in order to link it to the Java backend. When it comes to the coding, this is the distinction that can be seen the clearest between the two.

There is also a strong connection between JSP and JSF, which stands for Java Server Faces and is a Java standard for the construction of model-view-controller web applications. JSP stands for Java Server Pages, while JSF is an acronym for Java Server Faces. Java Server Pages (JSP) and Java Server Faces (JSF) are both abbreviations for Java Server technologies. Java Server Faces (JSF), PrimeFaces, MyFaces, and Eclipse Mojarra are all examples of Java web frameworks. JSF is often known by its full name, Java Server Faces. JSF is the standard used in this business. On the other hand, JSP is a technology that is not only less difficult but also more well-established than a lot of its competitors.



One example of a web container is the Apache Tomcat programme, which was built by Apache. Users are given the chance to execute programmes that are constructed using Servlet and Java Server Pages and are based on web-based software. It is possible to use it in the function of an HTTP server if one so chooses. When compared to the performance of the web server that will be utilised, the Tomcat server has a much lower level of functionality. Due to the fact that it already incorporates its very own web server, it is possible to make use of it on its own as a solution that may be used independently. Additionally, it is feasible to utilise it in conjunction with other web servers, such as the Microsoft Internet Information Server, the Apache web server, or the Microsoft Personal Web Server.

MySQL is a relational database management system that may be downloaded free of charge and is licenced under an open-source model. Tables, each of which is made up of rows and columns, are used to structure the data storage in MySQL, much as they are used in other relational databases. Users are able to generate, manage, and control data, as well as query for information about it, by using Structured Query Language, which is abbreviated to SQL for short. The name MySQL is derived from the combination of the words "My," which is the name of the inventor of MySQL and the daughter of Michael Widenius, and the term "SQL."

Logistic regression is a kind of strategy for supervised learning categorization that is used to produce predictions on the likelihood of an outcome variable. Logistic regression is a form of approach for supervised learning categorization. This is accomplished by taking into consideration the connection that exists between the independent variables and the variables that are being analysed. The aim, also known as the dependent variable, has a dichotomous character, which indicates that there are only two distinct categories that may be selected as potential outcomes.

The dependent variable is of a binary type, which, to put it another way, means that its data can be represented as either a 1 (which stands for successful/yes) or a 0 (which stands for unsuccessful/no). Additionally, the dependent variable is of a binary type, which means that its data can be represented as either a 1 (which stands for successful/yes) or a 0.

In order for JSP to be put into action, it is necessary for a number of critical phases and components to be present. Within the context of a Java web application, we are going to get into the intricacies of implementing JSP in the next section.

Establishing the Conditions for Product Development The first thing that developers are responsible for doing is getting their development environments set up and ready to go. In order to complete this stage, it is often necessary to install the Java Development Kit (JDK) as well as an Integrated Development Environment (IDE) such as Eclipse or IntelliJ IDEA. These technologies provide the necessary architecture for developing, building, and delivering JSP files.

The Organisation of the Project The Model-View-Controller (MVC) design paradigm is used in the creation of the vast majority of Java web application projects. JSP files are used as the vehicle for implementing the View component of this design pattern. The View component is the one that is in charge of showing the user interface. The developers' project structure has to be designed in such a manner that the JSP views are kept separate from the business logic, which is referred to as the Model, and the request processing, which is referred to as the Controller. This must be done in order to meet the requirements of the project.

Production of JSP Files JSP files, which include a mixture of HTML, CSS, JavaScript, and JSP tags, are saved with the .jsp extension by programmers. JSP files may also be opened with the .jsp extension. The JSP file extension is used to store JSP files on your computer. JSP tags make it possible to include Java code into a page, which in turn makes it feasible to build dynamic content and communicate with backend components. JSP tags also make it easy to reuse code across several pages. JSP enables the utilisation of a broad assortment of tags, including scriptlets, expressions, declarations, directives, and custom tags, amongst others.

Tag Libraries Made Available by JSP JSP makes available standard tag libraries that simplify the vast majority of the steps necessary in the process of generating webpages. The Java Server Pages Standard Tag Library (JSTL), for example, provides tags for iterating over collections, manipulating dates, and



using logic based on conditions, amongst other things. The functionality of JSP is improved by tag libraries like as JSTL, which also help to promote the reusability of code.

Integration with the Java Backend JSP pages may obtain and process data by interacting with Java backend components such as servlets and JavaBeans. This allows for seamless integration with the Java platform. Java gives programmers the ability to define classes and methods that can handle business logic and link with databases and other external systems. These capabilities may be used to build applications. By employing scriptlets, expressions, or custom tags, it is possible to activate the Java code that is included inside JSP. This allows the production of dynamic content that is dependent on the data that is received from the backend.

Processing Requests: JSP pages are often accessible via the use of URLs. Web servers then direct the requests to the appropriate JSP file based on the mappings that have been set. The web container is the component that is accountable for handling incoming requests, doing the necessary processing on those requests, and generating dynamic answers. The JSP file is compiled into a servlet by the web container whenever it receives a request for the website.

Deployment: Once the JSP files and associated Java classes are ready, they need to be deployed on a web server or a Java EE container such as Apache Tomcat or JBoss. Deployment occurs once the JSP files and related Java classes are ready. When both the JSP files and the related Java classes are ready, the deployment process may begin. As part of the deployment process, the project is first compressed into a file known as a Web Application Archive (WAR), and then that WAR file is transferred to the server of your choosing. The server is responsible for the execution of JSP files; at the same time, the server makes these files available to clients over HTTP.

Testing and Debugging: Once the application has been deployed, developers should do comprehensive testing on it to ensure that the JSP pages are displaying correctly and connecting with the backend as they were designed to do. The testing process involves verifying a variety of aspects, including the functioning of a web application, its performance, and the overall user experience. During the phases of development and testing, debugging tools that are provided by the integrated development environment (IDE) or logging frameworks may be useful in discovering and correcting issues that have occurred during the course of development.

When the JSP-based web application is live on the internet, regular maintenance and software updates are obligatory in order to guarantee that it will continue to operate correctly. Fixing problems, adding new features, enhancing performance, or adapting to the shifting demands imposed by the firm are all examples of things that might fall under this category. Due to the fact that JSP is modular in design, it is feasible for developers to modify certain JSP files without those modifications having an effect on the operation of the programme as a whole.

Putting it succinctly, implementing JSP requires setting up the development environment, producing JSP files that include a combination of HTML and JSP tags, integrating Java backend components, managing request processing, deploying the application to a web server, and testing the functionality in great detail. By following to best practises, making use of the required tools and frameworks, and so on, it is possible to simplify the process of implementation and increase the possibility of successfully constructing dynamic and data-driven Java web applications using JSP. This may be achieved by streamlining the implementation process and improving the likelihood of success.

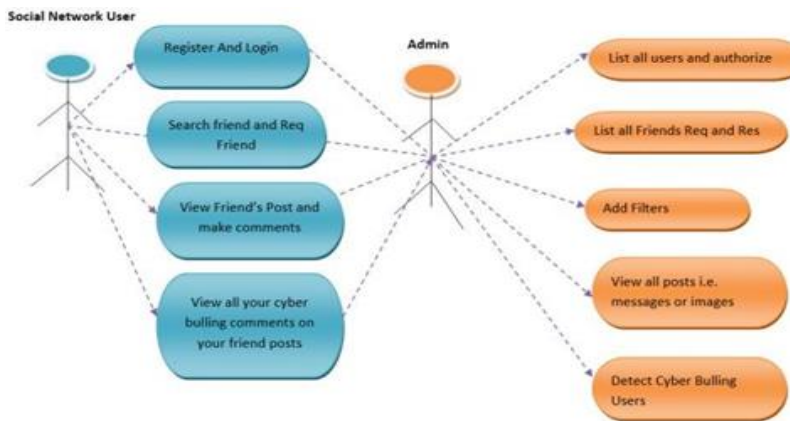


Fig. 1: Use Case Diagram

Fig. 1 shows the use case diagram for the implemented cyberbullying detection and how social network users and admins interact with the system.

Comment Posting Status..

This is an Cyber Bullying Post, So this post can't be submitted !!..

Type : Violence

Count : 1

Words : [kill]

=====

[Back](#)

Fig. 2: Cyberbullying Posts of Type Violence

Comment Posting Status..

This is an Cyber Bullying Post, So this post can't be submitted !!..

Type : Hate

Count : 1

Words : [bad]

[Back](#)

Fig. 3: Cyberbullying posts of type Hate

Adding Filters..

Select Filter : Select

Enter Word :

Back

Sidebar Menu

Home

Log Out

All Filter Words..

Violence	[kill]
Vulgur	[Stupid, Bulliah]
Offensive	[Fat, Rascal]
Hate	[bad, nonsense]
Sexual	[Booms, fuck]

Back

Fig. 4: Add Filter Section

Fig. 5 and Fig. 6 shows cyberbullying details of posts that were indulged in the cyberbullying activity.

ID	User Name	Post Name	Cyberbullying Type	Cyber Word	Count	Date and Time/div>
1	Neeraj	Iphone	Offensive	[kill]	1	29/07/2021 12:59:52
2	Aditya	Iphone	Hate	[bad]	1	29/07/2021 13:47:18
3	Aditya	Iphone	Sexual	[fuck]	1	29/07/2021 13:48:53
4	Aditya	Iphone	Sexual	[booms]	1	29/07/2021 13:49:33
5	Sarab	Laptop LG	Vulgur	[stupid]	1	29/07/2021 14:37:01
6	Rahul	Iphone	Hate	[nonsense]	1	29/07/2021 15:11:04

Fig. 5: Cyberbullying details 1

7	Sarab	Bike_Activa	Offensive	[rascal]	1	29/07/2021 15:12:12
8	Rahul	Iphone	Offensive	[kill]	1	29/07/2021 15:15:02
9	Rahul	Iphone	Vulgur	[stupid]	1	29/07/2021 15:15:02
10	Hercules	Bike_Activa	Violence	[kill]	1	08/08/2022 17:11:46
11	Hercules	Bike_Activa	Hate	[bad]	1	08/08/2022 17:14:45

Fig. 6: Cyberbullying details 2

Fig. 7 shows users that have said the words the filter considers cyberbullying. Clicking on post details shows Fig.

Sl No.	User Image	User Name	
1		Neeraj	Posts Details
2		Aditya	Posts Details
3		Sarab	Posts Details
4		Rahul	Posts Details
5		Hercules	Posts Details

Fig. 7: Cyber Bullying Users

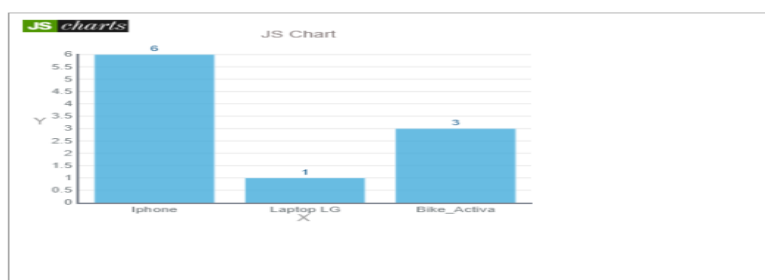
Cyber Bullying Comments by Neeraj ..

	Post Title	Iphone
	Created By	Sarab
	Date	29/07/2021 12:48:38
Type	Bullying Trace,(cyberbullying words,No.of Time commented)	
Offensive	[[kill]], Count=1	

[Back](#)

Fig. 8: Cyber Bullying Post Details

Cyberbullying Review Results..



[Back](#)

Fig. 9: Cyberbullying Chart Results

Fig. 9 shows chart results of cyberbullying posts.

IV.CONCLUSION

The technology that has been developed makes use of a filter known as the cyberbully filter in order to identify certain users and posts as instances of cyberbullying. The unsupervised machine learning classifier is what is used in order to arrive at these conclusions. When social media networks have access to this information, it is much simpler for them to detect instances of cyberbullying at an early UGC CARE Group-1



stage and to put in place bans and other preventive steps according to the severity of the cyberbullying post. In conclusion, the progression of technology, especially in the fields of machine learning and big data analytics, has created useful tools for identifying and combatting cyberbullying on social media platforms. These technologies can be found on platforms such as Facebook, Twitter, and Instagram. The creation of cyberbully filters and unsupervised machine learning classifiers has made it possible to identify persons and postings that are related with bullying behaviour in the online space. Social media networks are able to identify instances of cyberbullying at an early stage by utilising these technologies and implementing relevant measures, such as bans or preventative actions, depending on the severity of the cyberbullying post.

The use of these cutting-edge technology has enormous ramifications for the process of making the internet a more secure place to do business. It gives social media firms the ability to proactively handle incidences of cyberbullying and safeguard users from the negative repercussions of engaging in aggressive behaviour online. Platforms have the ability to lessen the potential damage inflicted on victims and develop an online community that is inclusive and supportive if they quickly recognise and react to incidents of cyberbullying. The battle against cyberbullying, on the other hand, is a continuing effort, which is something that has to be acknowledged. It is possible that cyberbullies may modify their methods as technology continues to advance; hence, it is essential for researchers and platform administrators to maintain vigilance and modify their approach appropriately. In addition, it is essential for scholars, legislators, and social media platforms to work together in order to create complete solutions that address the many facets of cyberbullying.

In conclusion, the integration of machine learning algorithms, big data analytics, and cyberbully filters has opened potential possibilities for addressing the problem of cyberbullying inside social media platforms. By using these tools and putting preventative measures into place, we may work towards the goal of creating a digital environment that is secure, welcoming, and free from the potentially damaging impacts of cyberbullying. In order to keep ahead of the trends of cyberbullying and to ensure the well-being of online communities, continued research and cooperation are going to be essential.

V. REFERENCES

- [1] M. A. Al-Garadi et al., "Predicting Cyberbullying on Social Media in the Big Data Era Using Machine Learning Algorithms: Review of Literature and Open Challenges," in *IEEE Access*, vol. 7, pp. 70701-70718, 2019, doi: 10.1109/ACCESS.2019.2918354.
- [2] B. A. H. Murshed, J. Abawajy, S. Mallappa, M. A. N. Saif and H. D. E. Al-Ariki, "DEA-RNN: A Hybrid Deep Learning Approach for Cyberbullying Detection in Twitter Social Media Platform," in *IEEE Access*, vol. 10, pp. 25857-25871, 2022, doi: 10.1109/ACCESS.2022.3153675.
- [3] A. Aggarwal, K. Maurya and A. Chaudhary, "Comparative Study for Predicting the Severity of Cyberbullying Across Multiple Social Media Platforms," 2020 4th International Conference on Intelligent Computing and Control Systems (ICICCS), 2020, pp. 871-877, doi: 10.1109/ICICCS48265.2020.9121046.
- [4] H. Hosseinmardi, R. I. Rafiq, R. Han, Q. Lv and S. Mishra, "Prediction of cyberbullying incidents in a media-based social network," 2016 IEEE/ACM International Conference on Advances in Social Networks Analysis and Mining (ASONAM), 2016, pp. 186-192, doi: 10.1109/ASONAM.2016.7752233.
- [5] Y. J. Foong and M. Oussalah, "Cyberbullying System Detection and Analysis," 2017 European Intelligence and Security Informatics Conference (EISIC), 2017, pp. 40-46, doi: 10.1109/EISIC.2017.43.
- [6] R. Sugandhi, A. Pande, S. Chawla, A. Agrawal and H. Bhagat, "Methods for detection of cyberbullying: A survey," 2015 15th International Conference on Intelligent Systems Design and Applications (ISDA), 2015, pp. 173-177, doi: 10.1109/ISDA.2015.7489220.
- [7] S. Parime and V. Suri, "Cyberbullying detection and prevention: Data mining and psychological perspective," 2014 International Conference on Circuits, Power and Computing Technologies [ICCPCT-2014], 2014, pp. 1541-1547, doi: 10.1109/ICCPCT.2014.7054943



- [8] W. Romsaiyud, K. na Nakornphanom, P. Prasertsilp, P. Nurarak and P. Konglerd, "Automated cyberbullying detection using clustering appearance patterns," 2017 9th International Conference on Knowledge and Smart Technology (KST), 2017, pp. 242-247, doi: 10.1109/KST.2017.7886127.
- [9] K. Reynolds, A. Kontostathis and L. Edwards, "Using Machine Learning to Detect Cyberbullying," 2011 10th International Conference on Machine Learning and Applications and Workshops, 2011, pp. 241 -244, doi: 10.1109/ICMLA.2011.152.
- [10] B. Haidar, M. Chamoun and F. Yamout, "Cyberbullying Detection: A Survey on Multilingual Techniques," 2016 European Modelling Symposium (EMS), 2016, pp. 165-171, doi: 10.1109/EMS.2016.037.
- [11] A. Kovačević, "Cyberbullying detection using web content mining," 2014 22nd Telecommunications Forum Telfor (TELFOR), 2014, pp. 939 -942, doi: 10.1109/TELFOR.2014.7034560.
- [12] N. Potha and M. Maragoudakis, "Cyberbullying Detection using Time Series Modeling," 2014 IEEE International Conference on Data Mining Workshop, 2014, pp. 373-382, doi: 10.1109/ICDMW.2014.170.
- [13] D. Zois, A. Kapodistria, M. Yao and C. Chelmis, "Optimal Online Cyberbullying Detection," 2018 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP), 2018, pp. 2017- 2021, doi: 10.1109/ICASSP.2018.8462092.
- [14] D. Mouheb, M. H. Abushamleh, M. H. Abushamleh, Z. A. Aghbari and I. Kamel, "Real-Time Detection of Cyberbullying in Arabic Twitter Streams," 2019 10th IFIP International Conference on New Technologies, Mobility and Security (NTMS), 2019, pp. 1-5, doi: 10.1109/NTMS.2019.8763808.
- [15] V. Jain, V. Kumar, V. Pal and D. K. Vishwakarma, "Detection of Cyberbullying on Social Media Using Machine learning," 2021 5th International Conference on Computing Methodologies and Communication (ICCMC), 2021, pp. 1091-1096, doi: 10.1109/ICCMC51019.2021.9418254.
- [16] K. Wang, Q. Xiong, C. Wu, M. Gao and Y. Yu, "Multi-modal cyberbullying detection on social networks," 2020 International Joint Conference on Neural Networks (IJCNN), 2020, pp. 1-8, doi: 10.1109/IJCNN48605.2020.9206663.
- [17] A. T. Aind, A. Ramnaney and D. Sethia, "Q-Bully: A Reinforcement Learning based Cyberbullying Detection Framework," 2020 International Conference for Emerging Technology (INCET), 2020, pp. 1-6, doi: 10.1109/INCET49848.2020.9154092.
- [18] A. Bozyiğit, S. Utku and E. Nasiboğlu, "Cyberbullying Detection by Using Artificial Neural Network Models," 2019 4th International Conference on Computer Science and Engineering (UBMK), 2019, pp. 520-524, doi: 10.1109/UBMK.2019.8907118.
- [19] J. Zhang, T. Otomo, L. Li and S. Nakajima, "Cyberbullying Detection on Twitter using Multiple Textual Features," 2019 IEEE 10th International Conference on Awareness Science and Technology (iCAST), 2019, pp. 1-6, doi: 10.1109/ICAwST.2019.8923186.
- [20] R. Pawar and R. R. Raje, "Multilingual Cyberbullying Detection System," 2019 IEEE International Conference on Electro Information Technology (EIT), 2019, pp. 040-044, doi: 10.1109/EIT.2019.8833846.