

# A Noval Method to Detect Cyber-Attacks in IOT Devices Dataset using DL

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**Abstract:** *In the modern era, the usage of internet has increased tremendously which in turn has led to the evolution of large amount of data. Cyber world has its own pros and cons. One of the alarming situations in web 4.0 is cyber bullying a type of cyber-crime. When the bullying occurs on line with the aid of technology it is known as cyber bullying. This research paper have surveyed the work done by 30 different researchers on cyber bullying, and elaborated on different methodologies adopted by them for the detection of bullying. Cybercrimes involve all the crimes where internet is used as an access medium and committed through some electronic device such as computers and mobile phones. Unavailability of datasets, hidden identity of predators and the privacy of the victims are the main factors for limiting the past research in cyberbullying detection. Considering these factors, an effective text mining approach using machine learning algorithms is proposed to proactively detect bullying text. The dataset collected from myspace.com and Preverted-Justice.com has been used to evaluate the system's performance. Three types of feature namely textual, behavioral and demographic features are extracted from the dataset as compared to earlier study over the same dataset where only textual features were considered. Textual features include certain bullying words that if exists within the text may lead to a true outcome for cyberbullying. Personality trait features are extracted for the user if it is involving once in bullying may bully in future too. While demographic features extracted from dataset include age, gender and location. The system is evaluated through different performance measures for both classifiers used and the performance of Support Vector Machine classifier is found better than the Bernoulli NB with an overall 87.14accuracy.*

**Keywords:** Bullying, Cyber, detection, feature extraction, logistic regression classifier, IOT, and attacks

## I. INTRODUCTION

### Overview

Across the globe due to the tremendous increase in the availability of data services, addiction of social media among the society has increased proportionally. Just like other countries, India has also witnessed a drastic rise in the cyber bullying. In this era of web 4.0 where people live in digital and online platforms, it is very difficult to protect the society from the alarming rise in cyber-crime. It has been surveyed that the major victims of cyber bullying are adolescents. Different cyber bullying attacks that are performed by attacker are: (1) Sending or posting hateful or abusive comments with an intention to harm the character of an individual (2) Posting an inappropriate image or video. (3) Creation of a false or improper website.(4) Issuing online threats that cause a person to kill themselves or injure another person. (5) Triggering online religious, racial, ethnic or political hatred by posting hate comments or videos.

### Motivation

The main Motivation is to Avoid Cyber Bullying and save student or any human life. Although some users indicate they are being sarcastic, most of them do not. Therefore, it might be indispensable to and a way to automatically detect any sarcastic messages Cyber bullying is threatening and destructive act which may result in suicide attempts or

negative impact can cause life-long harms to the victims. The detection of Cyber bullying can be considered as a classification problem. An online post can be classified as a bullying post or normal post. We will develop a system by applying different machine learning methods to better detect Cyber bullying and improve performance. For example, the Support Vector Machine(SVM), Forest Classifier **Objective** To The objective of the system is to reveal, analyse and stop cyber bullying in social media applications. To identify the occurrence of cyber bullying activity in social media platform which helps the government to yield force before many end-users enhance their Target of cyber bullying. To The system is to give alert message like to warn them, and to identify short hand text and human aggressive behavior on the comment sections. To Also to generate a report which contains the details of bully, and to keep track of count and also by blocking that person along his comment without letting it reach to victim.

## II. LITERATURE REVIEW

"Rapid Cyber-bullying detection method using Compact BERT Models": This paper use various compact BERT models and fine-tune them with hate-speech data. We incorporate Focal Loss function to handle class imbalance in the data. Using this approach, we were able to achieve state-of-the-art results of 0.91 precision, 0.92 recall and 0.91 F1-score on the hate-speech dataset. Additionally, using our transfer learning pipeline, we show that the more compact BERT models are significantly faster in detection and are suitable for real-time applications of cyberbullying detection.

"Review of Machine Learning methods for Identification of Cyberbullying in Social Media": In this paper modern era, the usage of internet has increased tremendously which in turn has led to the evolution of large amount of data. Cyber world has its own pros and cons. One of the alarming situations in web 4.0 is cyber bullying a type of cyber-crime. When the bullying occurs on line with the aid of technology it is known as cyber bullying. This research paper have surveyed the work done by 30 different researchers on cyber bullying, and elaborated on different methodologies adopted by them for the detection of bullying, and how you protect the society from online evil act of cyber bullying.

"Identification of Potential Cyber Bullying Tweets using Hybrid Approach in Sentiment Analysis": This paper rise of the Internet, cyber bullying is becoming more and more widespread. Cyber bullying has resulted in such disastrous consequences that there is a pressing need to detect it. The aim of this study is to do the same by using sentiment analysis. We perform cyber bullying detection using a novice approach on Tweets using Natural Language Processing and Machine Learning techniques. After processing a tweet, it can be flagged down if the tweet is a potential cyber bullying threat.

"Analysis of Cyber Aggression and Cyber-bullying in Social Networking": This paper considers Ask.fm, a social networking site where users create profiles and can send each other questions, and analyses aggressive user behavior that may potentially lead to cyber-bullying incidents. We hypothesize that anonymity is a primary cause of such aggressive user behavior and examine how anonymous and non-anonymous users behave in social networking. We collected data from Ask.fm and analysed questions posted by anonymous and non-anonymous users and answers posted by non-anonymous users. Analysis of the collected data shows that anonymous users exhibit more aggressive behavior than non-anonymous users. Analysis also shows that users become more aggressive in answering aggressive anonymous questions than aggressive non-anonymous questions.

## III. SYSTEM ANALYSIS

### Module

#### 1. Admin

In this module, the Admin has to log in by using valid user name and password. After login successful he can do some operations such as View All Users and Authorize, View All E-Commerce Website and Authorize, View All Products and Reviews, View All Products Early Reviews, View All Keyword Search Details, View All Products Search Ratio, View All Keyword Search Results, View All Product Review Rank Results.

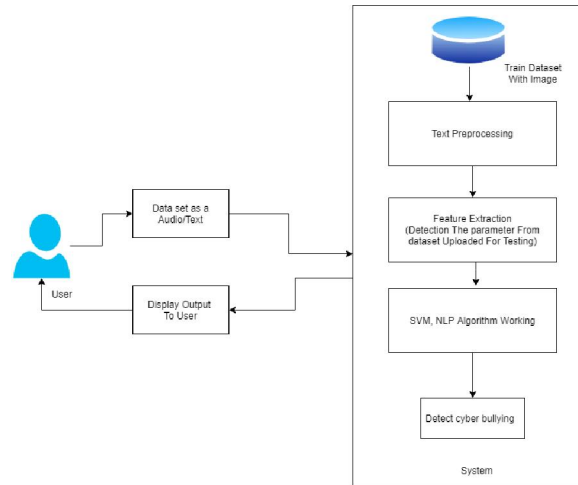


Figure: System Architecture

**View and Authorize Users**

In this module, the admin can view the list of users who all registered. In this, the admin can view the user's details such as, user name, email, address and admin authorizes the users.

**View Charts Results**

View All Products Search Ratio, View All Keyword Search Results, View All Product Review Rank Results.

**Ecommerce User**

In this module, there are n numbers of users are present. User should register before doing any operations. Once user registers, their details will be stored to the database. After registration successful, he has to login by using authorized user name and password. Once Login is successful user will do some operations like Add Products, View All Products with reviews, View All Early Product's reviews, View All Purchased Transactions.

**End User**

In this module, there are n numbers of users are present. User should register before doing any operations. Once user registers, their details will be stored to the database. After registration successful, he has to login by using authorized user name and password. Once Login is successful user will do some operations like Manage Account, Search Products by keyword and Purchase, View Your Search Transactions, View.

**Data Flow:**

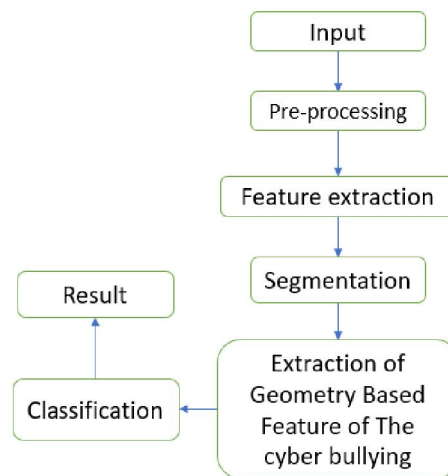


Figure: Data Flow Diagram

In DFD we show actual input and actual output of system input of our system is text or image and output is rumor detected in DFD we present operation of user as well as admin.

**System User:**

Admin is automatically registered by the system. The administrator must log in to the system and carry out any desired tasks. For normal users to access the system, they must first register and then log in

**Dataset:**

The user must gather the cyber bullying dataset. The dataset has thousands of rows and close to the model is trained using datasets of this type. The bullying of dataset.

**Pre-processing:**

Pre-processing of the dataset is done after it has been gathered. Pre-processing, put simply, is the process of transforming unprocessed data into a format that a machine can understand and use to build a machine learning model.

**Features:**

editor with syntax highlighting, introspection, code completion Support for multiple I Python consoles The ability to explore and edit variables from a GUI A Help pane able to retrieve and render rich text documentation on functions, classes and methods automatically or on-demand A debugger linked to IPDB, for step-by-step execution Static code analysis, powered by Pylint A run-time Profiler, to benchmark code Project support, allowing work on multiple development efforts simultaneously A built-in file explorer, for interacting with the filesystem and managing projects A "Find in Files" feature, allowing full regular expression search over a specified scope An online help browser, allowing users to search and view Python and package documentation inside the IDE A history log, recording every user command entered in each console An internal console, allowing for introspection and control over Spyder's own operation

**IV. SOFTWARE INFORMATION**

Python is an interpreted, high-level and general-purpose programming language. Created by Guido van Rossum and first released in 1991, Python's design philosophy emphasizes code readability with its notable use of significant whitespace. Its language constructs and object-oriented approach aim to help programmers write clear, logical code for small and large-scale projects. Python is dynamically typed and garbage-collected. It supports multiple programming paradigms, including structured (particularly, procedural), object-oriented and functional programming. Python is often described as a "batteries included" language due to its comprehensive standard library. Python was created in the late 1980s as a successor to the ABC language. Python 2.0, released in 2000, introduced features like list comprehensions and a garbage collection system with reference counting. Python 3.0, released in 2008, was a major revision of the language that is not completely backward-compatible, and much Python 2 code does not run unmodified on Python 3. The Python 2 language was officially discontinued in 2020 (first planned for 2015), and "Python 2.7.18 is the last Python 2.7 release and therefore the last Python2 release." [30] No more security patches or other improvements will be released for it. With Python 2's end-of-life, only Python 3.6.x and later are supported.

Python interpreters are available for many operating systems. A global community of programmers develops and maintains CPython, a free and open-source reference implementation. A non-profit organization, the Python Software Foundation, manages and directs resources for Python and CPython development.

Python was conceived in the late 1980s by Guido van Rossum at Centrum Wiskunde Informatica (CWI) in the Netherlands as a successor to the ABC language (itself inspired by SETL), capable of exception handling and interfacing with the Amoeba operating system. Its implementation began in December 1989. Van Rossum shouldered sole responsibility for the project, as the lead developer, until 12 July 2018, when he announced his "permanent vacation" from his responsibilities as Python's Benevolent Dictator For Life, a title the Python community bestowed upon him to reflect his long-term commitment as the project's chief decision-maker. He now shares his leadership as a member of a five-person steering council. In January 2019, active Python core developers elected Brett Cannon, Nick Coghlan, Barry Warsaw, Carol Willing and Van Rossum to a five-member "Steering Council" to lead the project.

**Anaconda:** Anaconda is a free and open-source distribution of the Python and R programming languages for scientific computing (data science, machine learning applications, large-scale data processing, predictive analytics, etc.), that aims to simplify package management and deployment. The distribution includes data science packages suitable for

Windows, Linux, and macOS. It is developed and maintained by Anaconda, Inc., which was founded by Peter Wang and Travis Oliphant in 2012. As an Anaconda, Inc. product, it is also known as Anaconda Distribution or Anaconda Individual Edition, while other products from the company are Anaconda Team Edition and Anaconda Enterprise Edition, both of which are not free.

Package versions in Anaconda are managed by the package management system conda. This package manager was spun out as a separate open-source package as it ended up being useful on its own and for other things than Python. There is also a small, bootstrap version of Anaconda called Miniconda, which includes only conda, Python, the packages they depend on, and a small number of other packages. Anaconda distribution comes with over 250 packages automatically installed, and over 7,500 additional open-source packages can be installed from PyPI as well as the conda package and virtual environment manager. It also includes a GUI, Anaconda Navigator, as a graphical alternative to the command line interface (CLI). The big difference between conda and the pip package manager is in how package dependencies are managed, which is a significant challenge for Python data science and the reason conda exists. When pip installs a package, it automatically installs any dependent Python packages without checking if these conflict with previously installed packages [citation needed]. It will install a package and any of its dependencies regardless of the state of the existing installation [citation needed]. Because of this, a user with a working installation of, for example, Google Tensorflow, can find that it stops working having used pip to install a different package that requires a different version of the dependent numpy library than the one used by Tensorflow. In some cases, the package may appear to work but produce different results in detail.

In contrast, conda analyses the current environment including everything currently installed, and, together with any version limitations specified (e.g. the user may wish to have Tensorflow version 2.0 or higher), works out how to install a compatible set of dependencies, and shows a warning if this cannot be done.

Open source packages can be individually installed from the Anaconda repository, Anaconda Cloud (anaconda.org), or the user's own private repository or mirror, using the conda install command. Anaconda, Inc. compiles and builds the packages available in the Anaconda repository itself, and provides binaries for Windows 32/64 bit, Linux 64 bit and MacOS 64-bit. Anything available on PyPI may be installed into a conda environment using pip, and conda will keep track of what it has installed itself and what pip has installed.

Custom packages can be made using the conda build command, and can be shared with others by uploading them to Anaconda Cloud, PyPI or other repositories.

The default installation of Anaconda2 includes Python 2.7 and Anaconda3 includes Python 3.7. However, it is possible to create new environments that include any version of Python packaged with conda

### **Spyder**

Spyder is a free and open source scientific environment written in Python, for Python, and designed by and for scientists, engineers and data analysts. It features a unique combination of the advanced editing, analysis, debugging, and profiling functionality of a comprehensive development tool with the data exploration, interactive execution, deep inspection, and beautiful visualization capabilities of a scientific package.

## **V. CONCLUSION**

In this work, a system is proposed which detects on English as well as on Hindi tweets in Twitter. Cyber bullying is very dependent and highly contextual; therefore, sentiment and other contextual clues to help detect the Cyber bullying. The system uses sarcastic tweets, 9,104 tweets containing Cyber bullying, and not dataset. The system uses the LR Algorithm. The approach has shown good results and it is observed that LR classifier has more accuracy than other classifier. All patterns for sarcastic detection are not covered in the extracted patterns. From the survey it can be concluded that the traditional machine learning algorithms are incapable of handling the enormous amount of data being generated in Web 4.0 moreover the cyber bullying content cannot be detected accurately. Recently Deep learning techniques, NLP, deep recurrent neural network, CNN, stacked auto-encoder, has gained the attention of many researchers. Future work can target on usage of these deep learning techniques for precise detection of cyberbullying in social media. Lot of research is being conducted in the field of cyberbullying. It is an emerging issue which needs to be addressed in Web 4.0. After reviewing the 30 research papers it found that there is a lack of proper dataset, collecting huge dataset is a major challenge, integrating social, contextual, sentiment features can improve the accuracy of

detection of bullying content. For future work, data from multiple social media platforms can be considered, apart from text, image, video must be taken into account for experimentation.

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