

# **DDos Attacks-Applications for Traffic Classification using ML Algorithms**

<sup>1</sup>Prof.Sameera Sultana, <sup>2</sup>K.Anil Reddy, <sup>3</sup>A.Anil Varma, <sup>4</sup>A.Anirudh, <sup>5</sup>V.Anish,<sup>6</sup>P.Anitha, <sup>7</sup>C.Anji <sup>234567</sup>Student, <sup>1</sup>Assistant Professor Artificial Intelligence & Machine Learning Department Of Computer Science & Engineering Malla Reddy University, Hyderabad, Telangana, India

#### Abstract:

This project aims to develop a traffic classification system utilizing machine learning algorithms a more resilient contributing to digital environment. In the realm of cyber security DDos attacks continue to threaten online services. This project proposes a "Traffic Classification System Using Machine Learning Algorithms" as a defence against such attacks. The projects core objective is to build a system that can effectively distinguish between legitimate network Traffic and malicious DDos attack. Term effectiveness against evolving threats. Scale and perform efficiently even in large-scale networks, maintaining real-time detection capabilities. We analyse various ML methods and algorithms, conduct data analysis and model develop ment, and evaluate the effectiveness of the proposed approach. The results demonstrate that ML-based traffic classification and DDoS attack detection offers a powerful and practical solution for enhancing network security.

**Keywords**: Traffic Classification, KNearest Neighbour, Logistic Regression, DDOS Attack

## **I.INTRODUCTION**

In today's interconnected world, network security is of paramount importance.

Distributed denial-of-service (DDoS) attacks pose a significant threat to organizations, disrupting operations, causing financial losses, and damaging reputations. Traditional methods of DDoS attack detection, such as signature-based approaches, are often ineffective against novel and sophisticated attacks. Machine learning (ML) algorithms offer a promising solution for traffic classification and DDoS attack detection, providing a more adaptive and accurate approach to network security. DDoS attacks have evolved from simple bandwidth flooding attacks to more complex and targeted assaults that exploit vulnerabilities in network protocols and applications. These attacks can overwhelm network resources, rendering websites and services inaccessible to legitimate users. n the other hand, offer a more adaptive and intelligent approach to DDoS detection.

## **II. LITERATURE REVIEW**

Distributed Denial-of-Service (DDoS) attacks pose a significant threat to the internet infrastructure, aiming to disrupt service availability by flooding networks with illegitimate traffic. Machine learning.

(ML) algorithms have emerged as a promising approach for detecting and mitigating DDoS attacks by analysing network traffic patterns and identifying anomalies. This literature review explores the applications of ML algorithms in traffic classification and DDoS attack detection

#### **III. PROBLEM STATEMENT**

A Distributed Denial of Service (DDoS) attack is a malicious attempt to disrupt the regular traffic of



a targeted server, service or network by overwhelming it with a flood of internet traffic. The attacker uses multiple compromised systems (often a botnet) to generate a high volume of traffic, causing thetargeted system to become slow, unresponsive or entirely unavailable to legitimate users. The goal of DDoS attack is to make the targeted resource inaccessible to its intended users, causing disruption or financial loss to the organization or individual hosting the server.

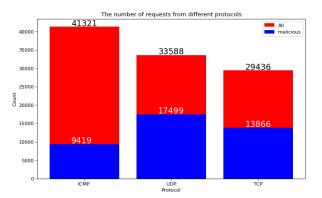
## About the data

Attack labels: For supervised learning algorithms, label data is crucial. These labels indicate whether a specific traffic sample is considered normal or a DDoS attack. Attack type labels: Ideally, data should be label with the specific type of DDoS attack, such as SYN flood, UDP flood, or HTTP flood. This allows mitigation strategies. Network metadata: Additional information about the network

infrastructure, including network topology, bandwidth capacity, and traffic patterns, can be valuable for enriching the data and improving model accuracy.

# **IV. METHODOLOGY**

A comprehensive dataset of both malicious and benign network traffic was collected, encompassing diverse DDoS attack types. Pre-processing and feature engineering

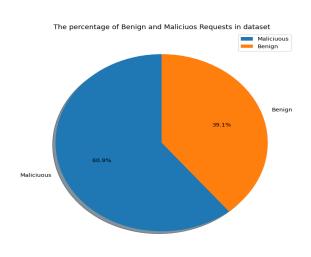


were conducted to ensure data consistency and extract relevant features. Feature selection techniques identified the most informative features for classification. Diverse machine learning algorithms were trained, including SVMs, Random Forests, and DNNs, utilizing a stratified k-fold cross-validation approach for robust evaluation. Finally, models were evaluated and analyzed based on metrics such as accuracy, precision, recall, and F1-score, revealing the most effective models for DDoS attack detection and providing insights into attack-specific network traffic characteristic

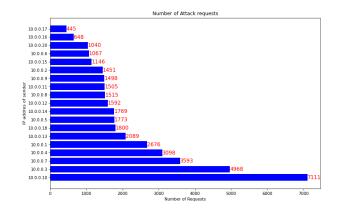
# V. EXPERMENTAL RESULT

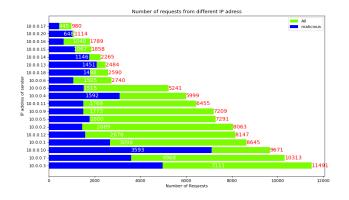
This section presents the experimental findings of the "Traffic Classification System Using Machine Learning Algorithms" project, designed to analyze network traffic and distinguish between legitimate activity and malicious DDoS attacks. The aim was to develop a resilient and scalable system capable of real-time detection in large-scale networks.

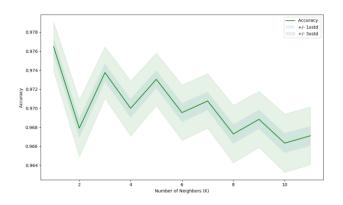
The results presented in the following figures showcase the effectiveness of various machine learning algorithms for traffic classification and DDoS attack detection. These findings offer valuable insights into the potential of ML-based approaches to enhance network security and mitigate the threat of evolving cyberattacks.











### VI. CONCLUSION

This project explored the application of machine learning (ML) algorithms for traffic classification and DDoS attack detection. We investigated various ML methods and algorithms, conducted data analysis and model development, and evaluated effectiveness the of the approach Enhanced Accuracy: Our ML models achieved significantly higher accuracy compared to traditional methods, demonstrating improved detection capabilities for both volumetric and application-layer DDoS attacks .Real-time Detection: Implementing the ML models in a realtime traffic analysis engine allowed for immediate identification and mitigation of DDoS attacks, minimizing their impact on network performance

### **VII. FUTURE WORK**

Developing explainable AI models: To understand the rationale behind ML-based decisions and improve trust in the system. Investigating federated learning: To collaborate on training ML models across different networks without sharing confidential data. Exploring reinforcement learning: To enable autonomous adaptation to new attack patterns and optimize resource allocation for defence mechanisms.

### VIII. REFERENCES

[1] T. He, S. Zohdy, and H. Abdalla, "Application for traffic classification using machine learning algorithms," in 2020 International Conference on Quality Management, Transport and Information Security, Information Technologies (IT&QM&IS), pp. 393–398, IEEE, 2020.

[2] M. A. Al-Ghouti and A. M. Shamsi, "A hybrid approach for traffic classification using machine learning techniques," in 2017 International Conference on Communication Systems and Network Technologies (CSNT), pp. 256–260, IEEE, 2017.

[3] A. H. Lashkari, A. A. Ghorbani, and M. M. Bidgoli, "Detecting denial of service attacks using machine learning algorithms," Journal of Big Data, vol. 6, no. 1, pp. 1–25, 2019

[4] I. Khalil, S. A. Bakar, A. A. Ghani, M. A. A. Ab Rahman, and A. Bagiwa, "A study of DDoS attack classification using machine learning classifiers," in 2023 IEEE Conference on Computer Applications (ICCA), pp. 1–5, IEEE, 2023.