

# Detection of DDoS Attack using Machine Learning: A Review

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**Abstract:** *Distributed Denial of Service, or DDOS attacks. A DDoS attack is a very disruptive form of attack, enabling the service to be rendered unsafe in some organizations that provide them with a connection to the internet community. This kind of threat keeps getting complicated and will most likely increase in number from one moment to the next. Thus, it becomes quite difficult to detect or devise a larger defense against this kind of threat. Hence, sophisticated intrusion detection systems (IDS) need to be developed, which will classify and identify abnormal internet traffic behavior.*

*More complicated and personal, however, they are advancing their number from one day to another, and detecting and defending against such threats is difficult. Thus, it demands decent intrusion detection system programming to classify and identify abnormal internet traffic behaviors.*

**Keywords:** Cyber Attack Detection, DDoS Attack Prevention, Intrusion Detection Systems, Machine Learning, Network Security,

## I. INTRODUCTION

Internet services are become a component of both individuals' and businesses' daily need. The increasing demand for network services has led to a rise in attacks by network intruders who aim to prevent genuine users from using the services. DDOS attacks [1] are those that interrupt or cause network programs to operate more slowly. Attackers use the millions of publicly accessible computer systems on the internet to launch a denial-of-service assault. As a result, servers frequently ignore responses from authorized users and remain occupied with requests brought about by assaults. The most well-known websites are those of banks, social media, and academic institutions. To protect the important data and services from those attackers, it is therefore vital to deploy one or more security solutions in a computer network, such as firewall intrusion detection systems (IDS) or anti-virus software.



Figure 1: DDoS Attack



One common way to deal with DDoS attacks is to use an intrusion detection system (IDS) [2]. An IDS is a component that guarantees the availability, confidentiality, and integrity of Web services and computer network resources. A system that uses machine learning algorithms is able to recognize and categories DDoS attacks, hence removing the incursion. It is difficult to attain complete performance accuracy in terms of attack detection and categorization, nevertheless.

## II. TYPES OF DDoS ATTACK

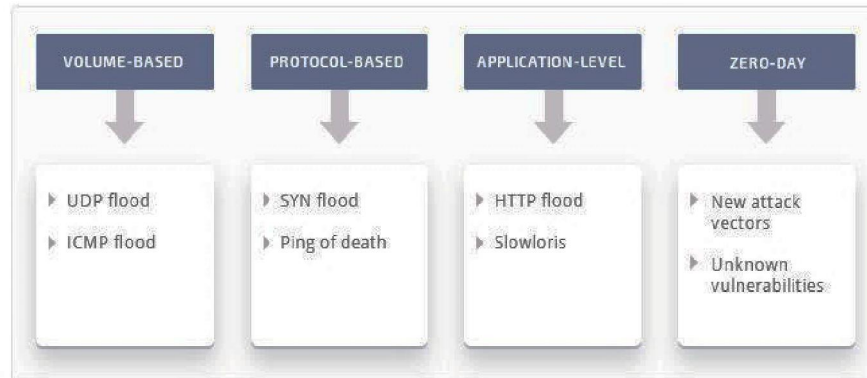


Figure 2: Types of DDoS Attack

### UDP Flood:

A UDP flood [3] is a type of DoS volume attack in which the attacker uses IP packets containing UDP packets to assault and share the host's random ports. The way the hosts look for apps linked to certain datagrams during this attack is explained in Figure 3 below. The host returns an "Unreachable Destination" packet to the sender if nothing is discovered. The network will become completely unresponsive to the much-needed legitimate traffic as a result of this torrent bombardment.

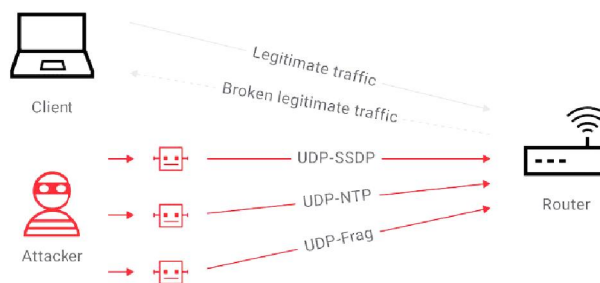


Figure 3: UDP Flood

### ICMP(PING) FLOOD:

The most frequent type of assault is called a "ping flood," or "ICMP flood," in which the attacker bombards the victim's equipment with ICMP echo requests, or "pings," forcing it to shut down. The ICMP flood attack, which involves flooding the victim's network with request packets while knowing that the system will respond with as many reply



packets as it can, will be described in the figure. Hopeful and fearful are the file kinds used to capture a target down for ICMP [7] and requests utilizing proprietary software or code

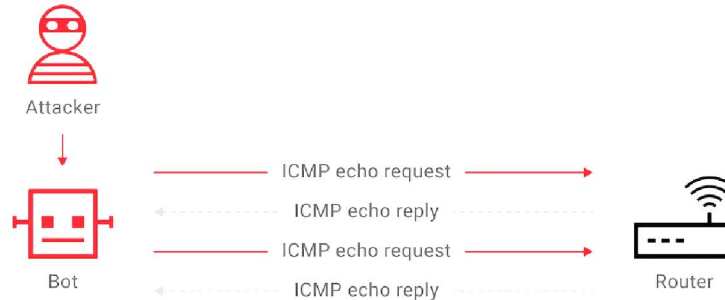


Figure 4: ICMP Flood

**SMURF ATTACK:**

This is another type of DDoS [9] assault in which a large number of Internet Control Message Protocol (ICMP) packets are transmitted to a computer network via an IP broadcast address, primarily utilizing the victim's spoof source IP. The majority of devices on a network will automatically reply by responding to the source IP address, as seen in the picture below. Traffic can overwhelm the attacker's computer if there are more systems than the size of the packet being received and responding to it.

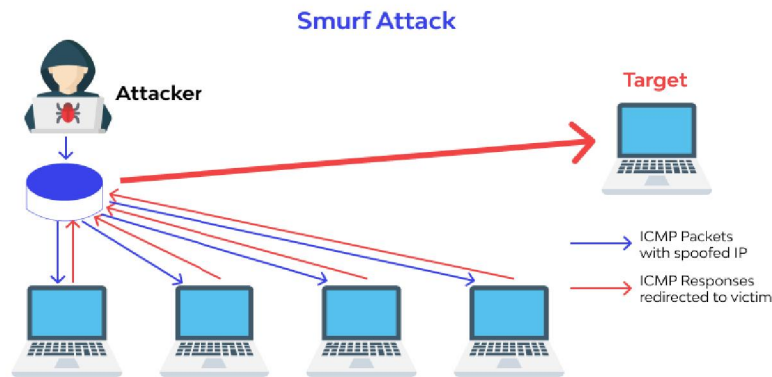


Figure 5: SMURF Attack

**HTTP FLOOD ATTACK:**

As seen in Figure 6, an HTTP Flood is a type of distributed volumetric Denial-of-Service attack. Its purpose is to overload a server with HTTP requests. Specific requests from actual users would experience denial-of-service whenever the target also becomes overloaded with queries and is unable to respond to normal traffic.



### HTTP Flood attack

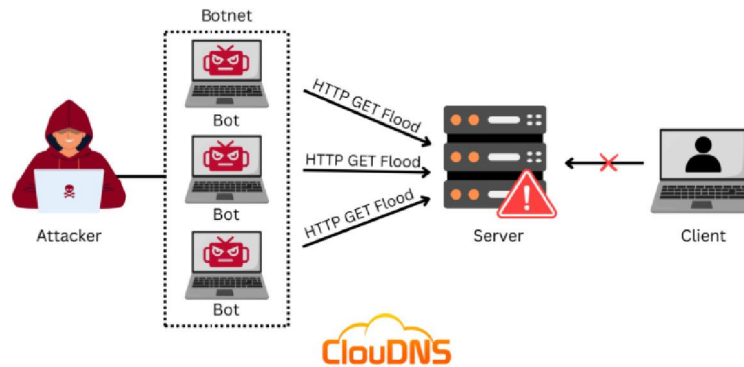


Figure 6: HTTP Flood Attack

### III. MACHINE LEARNING METHODS RELATED TO DDOS ATTACK DETECTION

Signature-based IDS is typically a human operation that would take long hours of testing developing and deploying a signature and creating another signature for unknown attacks too. It becomes very much human-less. Anomaly-based IDS evolved from Machine Learning languages and solves the problem by providing such a framework that it will learn from data and predict unknown states of information on what has been learned data.

#### Naïve Bayes

Naive Bayes is centered around the Bayesian classification framework. It is a straightforward and efficient method for establishing classifiers, which assign class labels to instances represented as feature value vectors. These class labels are derived from a specific finite set. Hidden Naive Bayes (HNB) [11] approach yields more reliable outcomes compared to the Standard Naive Bayes model. The HNB technique can be effectively employed to predict intrusion issues, such as Denial of Service (DOS) attacks, by leveraging strongly correlated dynamic characteristics and extensive network data stream capabilities.

The Naive Bayes algorithm is based on the principles of Bayesian classification. It is recognized as a simple and effective method for creating classifiers that assign class labels to cases represented as vectors of feature values, with these labels being derived from a limited set. The Hidden Naive Bayes (HNB) model demonstrates greater reliability than the Standard Naive Bayes model. This technique can be utilized to forecast intrusion threats, including DOS attacks, by taking advantage of highly correlated dynamic features and robust network data stream capabilities.

Naive Bayes operates on the Bayesian classification model, providing a straightforward and efficient approach to classifier development. This method assigns class labels to instances, which are defined as vectors of feature values, with labels derived from a finite set. Hidden Naive Bayes (HNB) model offers more dependable results than the Standard Naive Bayes model. The HNB technique is particularly useful for predicting intrusion issues, such as Denial of Service (DOS) attacks, by utilizing strongly linked dynamic characteristics and extensive capabilities in network data streaming.

#### Decision Tree:

One of the principal techniques that machine learning and data mining apply is the decision tree. Moreover, it may be used as a model of prediction in which findings about an object are mapped to inferences about the preferred value of the object. A decision tree can be applied to the investigation of decision data to draw out, in a very explicit and visual way, how decisions are made. In this method, a data set that is under study is developed. Therefore, if the new data element is given for classification the former dataset will classify it properly.



The Decision Tree Algorithm was used to detect denial-of-service attacks. Using classification techniques, data mining is used to detect Denial of Service assaults. There are two types in this particular approach: "normal" traffic and "anomalous" traffic. In addition to discussing the J48 decision tree algorithm's efficacy in Denial of Service assaults, the article will compare it to other rule-based algorithms like Decision Table. Md. Dewan. Farid et al. [3] presented an anomaly-based network intrusion detection learning method in their study that uses a decision tree approach to identify various types of intrusions and stops attacks from routine activities. The KDD99 dataset, which is primarily designed for network intrusion detection, is employed in this study.

#### **Artificial Neural Network:**

ChandrikaPalagiri demonstrated that a modelling network may recover a believable result for illustrating a neural network [12], specifically for a given attack. Researchers are also concentrating on a neural network that can identify choices quickly and in real time.

One notable choice for the base classifier in a publication by Madhav Kale et al. is Resilient Back-Propagation, or RBP. For the actual classification decision, the paper's focus was on improving the RBP classifier's efficiency by combining the outputs of classifiers using Neyman Pearson's cost reduction strategy. Detection accuracy and cost per sample were the two metrics used to gauge the effectiveness of the RBP Boost classification system.

The purpose of this research by Md Salem et al. [15] was to investigate if a firewall could actually track its traffic patterns in order to identify targeted denial of service attacks. This paper used statistical studies of firewall logs for a large network to analyse a baseline of the network. This study used statistical analysis of firewall logs for a broad network to determine a baseline.

Using the Holt-Winter and linear regression techniques, estimated traffic rates were computed to compare with the baseline. The analysis's findings were positive, indicating a widespread campaign in the network based on the departure from the expected rejected packet rates.

Author Mohammad Masoud Javidi et al [17] Proposed IDS that employs the supervised neuralnetwork for malicious DDOS in the NSLKDD database. The researcher used a signature-based methodology in the proposed IDS. IDSs are designed with a neural network that can detect various forms of DoS attacks and have a different IDS for each one to identify the particular attack.

#### **Support Vector Machine:**

T. Subbulakshmi [9] first proposed the concept of an SVM, which attracted a lot of interest from the machine learning research community. This approach uses examples with target values to classify and regress supervised learning techniques. The SVM algorithm uses a group of trained examples to generate a design that divides a technique into two classes based on which new instances are expected to closely fit into one of the two categories.

In DOS attacks; Packets are first grabbed from the network, and data is instantly fed into the RST. The RST-chosen feature sets would then be used in training and testing with the SVM model, etc. The findings were subsequently tested and showed to function through PCA). It shows that RST and SMV are well-capable of this task, and it increases the false positive rate for improving upward efficiency.

Subbulakshmi et al [9] have published a paper towards

Online network monitoring and activation of a security approach at a timewhenver doubtful activity is there. This approach helps identify both the non-spoofed IPs and spoofed IPs. The ESVM-based method has been employed in the work for spoof detection mechanisms throughspoofs identification in IP Hop Count Filtering has also been taken into consideration by the author to identify the spoofs in IP. These IPs are used to initiate the defense. The Lanchester Rule is applied to determine the attack force applied to trigger the defense mechanism.

#### **K-means clustering:**

K-Means clustering is a subset of unsupervised machine learning approaches [12] that enable partitioning of samples into K sets in a manner that maximizes the similarities within the samples in the set.





**How K-means Clustering Works:**

- **Initialization:** Choose K centroid either randomly or based on the average from each sample in the K groups.
- **Assignment:** It consists of grouping every sample to the K groups as per their closeness or distance to the K centroids.
- **Update:** Figure the average for every sample that was assigned to the K sections to enable determining the K centroids.
- **Repeat:** Observe step 2 and 3 continuously till the centroids stabilize or stop as per an alternative threshold.
- **Key Characteristics of K-means Clustering:**
- **Unsupervised learning:** Labeled data is not essential for its application [13].
- **Non-hierarchical:** Clusters don't conform to be children in other clusters.
- **Non-deterministic:** When set K is changed the clusters divided differ.
- **Advantages of K-means Clustering:**
- **Simple and efficient:** K-means is known for being quite simple and fast.
- **Effective for spherical clusters:** Essentially, K-means works for those clusters that are pinpointed by spherical shapes.
- **Disadvantages of K-means Clustering:**
- **Not suitable for non-spherical clusters:** Complex shapes tend to be hard for k to manage.
- **Requires choosing K:** When working with clusters, it is essential to pre-determine the cluster number K.

Real-World Applications of K-m

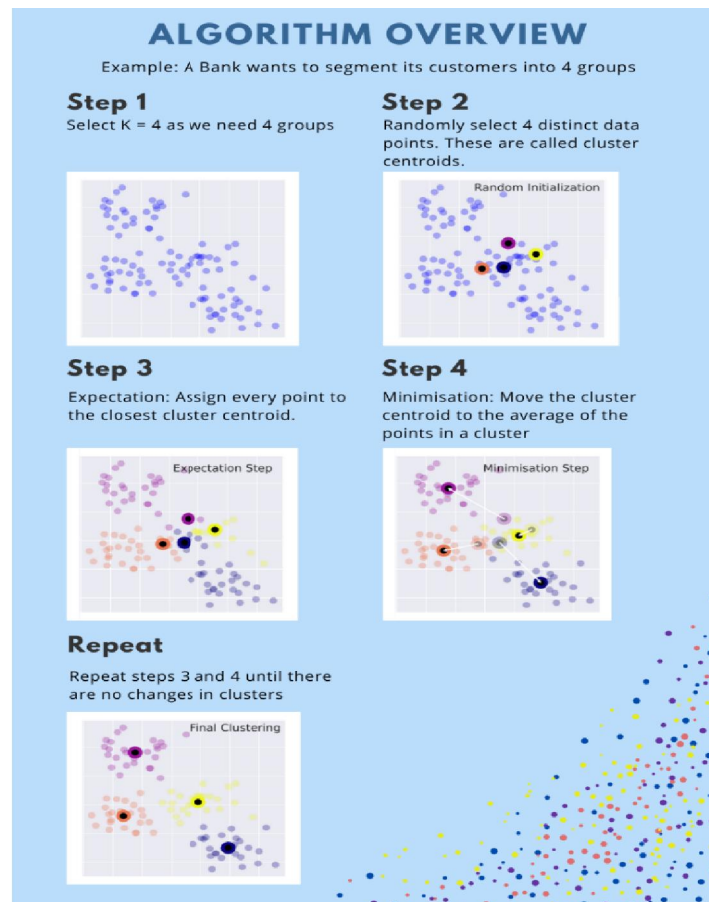


Figure7: Algorithm Overview

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#### **IV. CONCLUSION**

Finally, conclusion in the following comprehensive analysis, it is evident that web attacks pose serious risks, and IDS and IPS may not be sufficient to handle these new attacks that infect networks. Moreover, there is a necessity for machine learning approaches so that the severity of these attacks can be understood correctly and so that enterprises implement measures to prevent specific attacks.

#### **V. FUTURE ENHANCEMENT**

Recent trends in computer network attacks particularly the more advanced ones i.e., XXX, DDoS, Smurf and UDP flooding, as well as the newest types in as HTTP ontology should be adequate at this point. Thus, it will be possible to increase the level of protection provided by the firewall configuration.

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