



# A Literature Review on DDOS Attack Detection Using Artificial Neural Network

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**Abstract**—In internet technology, denial of service (DoS) assaults are a huge danger and one of the most difficult security issues. DDoS assaults, in particular, are a cause for worry due to their potential severity. A DDoS assault may swiftly deplete the victim's computational and communication capabilities in a short period of time with little or no prior notice. Due to the seriousness of the problem, a variety of defences have been devised.

**Keywords**-Distributed Denial of Service (DDoS), Cyber-attack, Confusion Matrix, feedforward neural network (FNN), Levenberg-Marquardt and Botnets.

## I. INTRODUCTION

A distributed denial of service (DDoS) attack is a malicious attempt to make an online service unavailable to users, usually by temporarily interrupting or suspending the services of its hosting server. A DDoS attack is launched from numerous compromised devices, often distributed globally in what is referred to as a botnet. It is distinct from other denial of service (DoS) attacks, in that it uses a single Internet-connected device (one network connection) to flood a target with malicious traffic

DDoS attacks are a serious issue on the Internet. The effect of DDoS attacks has been thoroughly documented in the computer network literature. Disruption is the primary goal of the DoS security services by restricting a computer or a network instead of attacking the service itself. This kind of thing an attack that seeks to make a network unusable normal service may be provided by focusing on either the network's ability to transmit data at high speeds. These assaults are successful. achieve their by the transmission of packets to a victim that the network or processing resources he has available are overtaxed. DDoS (Distributed Denial of Service) is a relatively new kind of attack. attacking Internet with a simple yet effective strategy resources. The many-to-one dimension is added by DDoS assaults. to the solution of the DOS issue D-DoS attacks often happen in stages, starting with hackers surveying or scanning for defenselessness or access points, First gaining access to a system and then

carrying out the assault in its entirety, whether the goal is to steal important data or to disable the computer systems. Automated instruction (ML), as well as deep learning (DL), have both been found to be completely effective in detecting DDoS attacks. Algorithms are, however, taught to detect only examples selected from the training set's distribution models [8]. As a result, individuals may perform in situations where they have never learned. The Open Set Recognition (OSR) challenge is about figuring out what one doesn't know. Because DDoS intrusion technologies develop, resulting in shifting traffic parameters, this issue has a significant influence on DDoS attack detection [9]. Malicious node attacks could not only prevent access to network resources but also result in major risks and harm.

A Denial of Service (DoS) attack, for example, prevents genuine end-users from using network resources by overflowing the target system by propagating widely and finally paralyzing it [10]. TCP/SYN Flood, Ping Flood, UDP Flood, and Distributed Denial of Service are all examples of DoS assaults (DDoS). A distributed denial of service (DDoS) assault occurs when a number of compromised malicious activities attack a single target. [25].

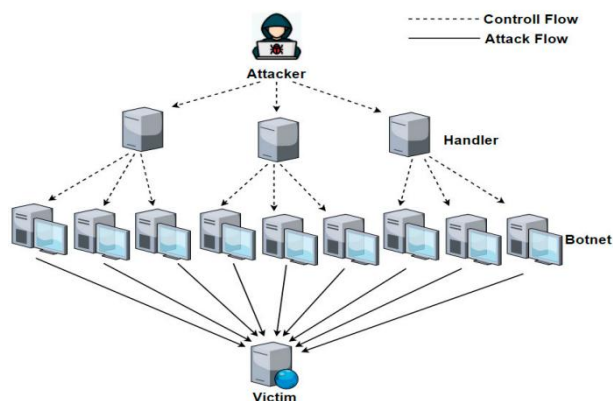


Fig. 1 Shows the D-DoS attacks using Bot Block diagram

In the above section I discuss the introduction part of proposed research work, discuss the cyber-attack and D-DoS attacks. In the section II discuss literature survey. Next section III discusses the classification DDOS attacks. Section IV discusses the DDOS attacks detection techniques. Last but not least discuss the conclusion in section V.

## II.LITERATURE SURVEY

*Kushwah et.al, (2021)*, In this research work authors proposed Cloud computing provides various types of resources in the form of services over the Internet. The availability of cloud services is crucial to the smooth functioning of this technology. Attackers can use DDoS attacks to disrupt the availability of cloud services. Authors presented work, a machine learning-based DDoS attack detection system for cloud computing is proposed. Self-adaptive evolutionary extreme learning machine (SaE-ELM) with crossover adaptation (SaE-ELM-Ca) is first established as an enhanced version of this device. The developed system is able to select the best mutation strategy, crossover rate, and crossover operator for a given situation. Authors determine the appropriate number of hidden neurons automatically. These features improve the learning capability of the model. That model is then used to construct the DDoS attack detection system. The performance of presented system was evaluated using state-of-the-art datasets NSL-KDD, ISCX IDS 2012, UNSW-NB15, and CICIDS 2017. It achieved the detection accuracy of 86.80%, 73.00%, 98.90%, 89.17%, and 99.99% with KD- DTest+, KDDTest-21, ISCX IDS 2012, UNSW-NB15, and CICIDS 2017 datasets, respectively [01].

*Jia et.al,(2020)*,In this research work investigator Virupakshar et.al, (2020), In this research work. The researchers reviewed 70 publications from high-profile journals. Some 47% of researchers utilised data theory approaches, 42percent utilised machine learning strategies, and 20% used Artificial Neural Networks to detect DDoS attacks in SDN. Additionally, they've provided details on various security mechanisms so that other researchers can better understand the current situation.

For SDN-enabled systems, operators remain a primary target for attackers [02].

*Dong et.al,(2019)*, In this research work DDoS attacks are becoming more prevalent in Analysts who say SDN and cloud computing environments pose the greatest risk. DDoS attacks and their detection in SDN and cloud computing are discussed at this moment by researchers. Because SDN could be a target of a DDoS attack, they look into how DDoS attacks are introduced on SDN as well as possible solutions. For DDoS attacks, researchers also discuss how to create exploratory conditions and then use simulation tools in SDN and cloud computing environments. Many unsolved issues in this area are examined, such as how to mitigate DDoS attacks in an SDN and cloud computing setting. Despite the never-ending research in this area, some issues remain unresolved. Future research should focus on this issue. They suggest the following lines of investigation for future work: Attacks against the SDN Even though SDN's unified regulation is its most attractive feature, it can also be a single point of failure when subjected to a Distributed Denial of Service (DDoS) attack. Traditional anomaly detection methods are having a hard time dealing with the growing number of DDoS attacks. As a result, SDN research is focusing on big data analysis and location advancements for DDoS attacks. SDN and NFV are mutually beneficial, but they are not dependent on each other. It is SDN's role to empower NFV. Because the logic for virtualization technology runs on a controller rather than on physical circuits, SDN helps to automate the network by allowing users to make approach-based decisions about how to coordinate system traffic flows [03].

*Dong et.al,(2019)*, In this research work DDoS attacks are becoming more prevalent in SDN and cloud computing environments, according to security analysts. At that point, researchers go over the various types of DDoS attacks as well as the tools used to detect them in SDN and cloud computing. In addition, since SDN might be a victim of DDoS attack, they survey the research work on introducing the DDoS attacks on SDN and how to solve this issue. they likewise talk about how to fabricate exploratory condition and use simulation instruments in SDN and cloud computing condition for DDoS attacks and identification. authors investigate some open research problems in this direction, for example, how to alleviate DDoS attacks in SDN and cloud computing environment. In spite of bottomless research that field, there are still some problems to be solved. That should be addressed in future research. Authors proposed some future research as pursues: DDoS attack in SDN: Albeit unified control is the real preferred advantage of SDN, it is additionally a solitary cause of failure also if it is targeted by Distributed Denial of Service(DDoS) attack. With the expansion in the DDoS attacks, traditional anomaly detection strategies face numerous troubles in alleviating the DDoS attacks. Hence, big information examination and location advancements

for DDoS attacks is a noteworthy region of research in SDN. SDN and NFV are mutually useful, yet are not subject to each other. SDN is an empowering influence for NFV. SDN contributes towards network automation that empowers approach based choices to coordinate system traffic flows, which is an engaging platform for network virtualization since control logic runs on a controller instead of on physical switches [09].

*Dayal, et. al (2017, January)*, In this research work, Researchers presented an attack model to identify and classify various DDoS attack scenarios in SDN. The hyena attack tool was used to carry out a variety of DDoS attacks in an SDN environment utilising a few of the most popular conventional DDoS attack methods. Despite the fact that volumetric attacks have a significant effect on the research plane, they do not have a significant impact on the controller. During the attack phase, the effect is clearly visible. Protocol exploitation threats, on the other hand, have little impact on network traffic. They focus on consuming other device resources, such as the TCAM, logical port, etc. Immediately after the attack, and immediately after the attack, controllers might be seriously impacted.. TCP SYN flood as well as HTTP flood attacks can in reality bring down the control system[10].

*Buragohain, Chaitanya, et.al. (IEEE 2016)*, In this research work, presented and tested the proposed

architecture of Flow Trapp. An SDN-based structure is provided for the detection and mitigation of both high- and low-rate DDoS attacks in data centers by this system. Incoming attack traffic can be classified using the optimization technique, which compares it to an application-specific tuple of genuine flow traffic. Remediation is used when a malicious user is discovered to transfer attack traffic regularly, rather than blocking the location at the outset. Architecture is implemented using SDN innovations such as OpenFlow and flow statistics collectors such as Flow To enhance the effectiveness of FlowTrApp, the OpenFlow controller, as well as the sFlow-RT application, can share the burden of detecting and mitigating DDoS attacks. Our method outperforms an existing QoS-based method in terms of performance[12].

*Zhao, T., Lo, et.al (2015, August)*.In this research work, Hadoop, as well as HBase, were used to design an effective DDoS detection method that can identify threats quickly. The first step was to create a Hadoop as well as HBase cluster to process a massive unorganised set of data. Once the neural network model for DDoS detection was created as well as six training data were used to train the neural network model, it was able to identify DDoS attacks. Three parts are used to demonstrate how well-suited the skilled neural network is for the task[16].

**Table 1 Comparison of Different Previous Methods**

S.No.	Ref./Year	Methods	Attack	Remark
1.	[1 ]/ 2021	Sae-Elm-Ca	Distributed denial-of-service (DDoS)	havegiventhewingstoCyber-PhysicalSystemadoption
2.	[2]/2020	CNN Algorithm	Distributeddenial-of-service(DDoS)	toclassifytheflowsintobenignonesoroneofthefourtypesofDDoS attacks.
3.	[03]/2019	Software-DefinedNetworks (SDN)	Denial of Service(DoS)	SDN decouplesthenetworkcontrolplane fromthedataplane.
4.	9/2019	Software Defined Networks (Sdn) Method	Denial of Service (DoS)	9/2019
5.	[10]/2017	Software-DefinedNetwork(SDN)	DDoSattacks	Toidentifyand classifyvarious possibilitiesofDDoSattacks

6.	[12]/2016	Artificial Neural Network	Detection DDoS attacks	To identify and detect abnormal traffic
7.	[16]/2015	Hadoop And HBase	Denial of Service (DDoS) attack	To process huge unstructured dataset

### III. DDOS ATTACK CLASSIFICATION

DDoS assaults may be divided into two categories: bandwidth depletion attacks and resource depletion attacks (Figure 3). An order to overwhelm the target, a bandwidth depletion attack is used. Traffic on a network that impedes the flow of lawful traffic the victim's system from receiving any unwanted traffic Bandwidth Flood attacks and amplification assaults are two types of attacks. Attacks depleting a person or organization's resources To stifle a victim system's ability to respond. This Protocol exploit attacks are a subcategory of kind of attack. In addition to erroneous packet assaults

Direct assaults and reflection attacks are the two main types of DDoS attacks. In the last part, we discussed direct assaults. A reflector is a kind of indirect attack that takes advantage of intermediate nodes to launch attacks. Any IP host that will return a packet if one is sent to it is a reflector.

DDoS attacks can be classified further as the primary target is to congest the network with a massive amount of the bandwidth. Utilization and it could cause the network abrupt to the victim network.

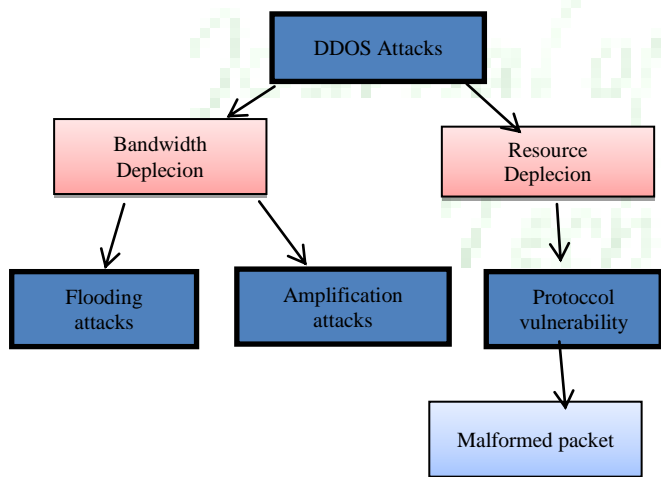


Fig. 2 Classification of DDoS attacks

DDoS attacks can be classified further as the primary target is to congest the network with a massive amount of the bandwidth Utilization and could cause the network abrupt-ion to the victim network. The secondary focus is on Network resource depletion. This means the

Attacker depletes the key components such as Memory, Device CPU, and so on. The attacker's intention behind this is to consume the available resources to the point where the service can't be responded to. Distributed Denial of Service (DDoS) attacks have been a real threat in many aspects of computer networks and distributed applications. The main objective of a DDoS attack is to bring down the services of a target using multiple sources that are distributed. For example, attackers can transfer thousands of packets to a victim to overwhelm its access bandwidth with illegitimate traffic, making online services unavailable. There are numerous denials of service (DoS) attack methods being used to degrade the performance or availability of targeted services on the Internet. Usually, these methods can be classified as challenges associated with the SDN at each layer of the framework, application, control, and infrastructure.

#### 1. Volume Based Attacks

Includes UDP floods, ICMP floods, and other spoofed-packet floods. The attack's goal is to saturate the bandwidth of the attacked site, and magnitude is measured in bits per second (Bps).

#### 2. Protocol Attacks

Includes SYN floods, fragmented packet attacks, Ping of Death, Smurf DDoS and more. This type of attack consumes actual server resources, or those of intermediate communication equipment, such as firewalls and load balancers, and is measured in packets per second (Pps).

**3. Application layer DDoS attacks** Application layer attacks use the software in a malicious way, aiming to exhaust resources to process any further requests. These attacks are generally harder to detect on the network level as they show no clear deviation from legitimate traffic. Since the isolation of applications or resources in SDN is not well solved, DDoS attacks on one application can also affect other applications. A common example is the HTTP flooding attacks.

#### 4. Control layer DDoS attacks

Controllers of SDN and their communications can be subjected to different types of attacks. Among the threats that can cause significant damages are the following: attacks on the control plane and communication between the controller and other networks components, eg, northbound API, southbound API, westbound API, or eastbound API. In addition, the controller can be considered a single point of failure and scalability that raises potential performance problems and unavailability of the control plane.

**5. Infrastructure layer DDoS attacks** Infrastructure layer DDoS attacks could potentially overload through two points: switches or by attacking the southbound API.<sup>22</sup> For example, huge traffic may be sent by an attacker to execute a DoS attack on the node by setting up a number of new and unknown flows infrastructure layer

#### IV. DDOS ATTACK DETECTION TECHNIQUES

DDoS attacks have been studied for a long time and the types of threats to them are mostly known.<sup>5</sup> However, SDN is a new architecture and the studies are at an early stage. In addition, SDN networks have distinct detection methods for different types of DDoS attacks.<sup>23</sup> These methods include entropy-based,<sup>7,24-26</sup> machine learning-based,<sup>10-13</sup> traffic pattern analysis,<sup>27</sup> connection rate,<sup>27,28</sup> and techniques that combine the use of the IDS and OpenFlow.<sup>29-31</sup>

**1. Entropy** The ability to measure randomness in packets arriving on a network makes entropy-based methods good candidates for the DDoS detection. The greater the randomness, the greater the entropy, and vice versa. Entropy-based methods depend on network feature distributions to detect anomalous network activities.<sup>24</sup> The presence of anomalies in an SDN network can be identified by adopting the use of predefined thresholds. In addition, probability distributions of various network features such as source IP address, destination IP address, and port numbers are used to calculate the entropy.<sup>32</sup>

**2. Machine learning** Machine learning-based methods employ techniques to detect anomalies in a network environment, these can be based on models, based on statistic and math, based on unsupervised machine learning algorithms, and based on supervised machine learning algorithms.<sup>33</sup> These algorithms take into account various network features and traffic characteristics to detect the presence of anomalies. In fact, any system that is built to detect any anomalies in the network catch the traffic on it and extract some kind of information from them, and the approach that uses machine learning is trying to catch the pattern of normal and abnormal traffic on the network, without the need to know the pattern itself

**3. Traffic pattern analysis** These techniques work on the assumptions that the infected hosts exhibit similar behavioral patterns that are different from benign hosts.<sup>34</sup> Therefore, it analyzes the traffic relating to the metrics of the attack pattern networks in order to identify the attacker or the target under attack. Patterns are observed as a result of a command that is sent to many members of the same botnet causing a similar behavior (eg, sending illegitimate packets, starting to scan).

**4. Connection rate** Bawany et al<sup>32</sup> define connection rate techniques as “the probability of a connection attempt being successful should be much higher

for a benign host than a malicious host.” Whenever the likelihood ratio for a host to exceed a certain threshold, it is declared as infected. These techniques are classified into two types: connection success ratio and connection rate, which refers to the number of connections instantiated within a certain window of time.

#### V. CONCLUSION

In this research work researcher presented different method used on DDOS attack detection a Bayesian regularization with cascaded feed forward network-based machine learning approach for D-DoS attack detection. The presented method shows better result as compare to CNN with LSTM based deep learning methods. Table II shows the comparison between presented method and previous methods.

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