

Machine Learning for Intrusion Detection Systems

Mr.B.Senthil Kumar¹, Dr.M.S.Josephine², Dr. V.Jeyabalaraja³

¹Research Scholar, Bharathiar University, Coimbatore, senthilkumar32@yahoo.com

²Professor, Dr.MGR Educational and Research Institute, josejbr@yahoo.com

³Professor, Velammal Engineering College, jeyabalaraja@gmail.com

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Abstract

In recent decade most of technologies are evolved and there security handling also improved. In which, IDS is the software which is used to detect unauthorized intruders in the network. Even though the highly secure devices and there security feature are developed day-by-day. The malicious hackers update their techniques to crack the security by identifying the vulnerability in the network. Lots of intrusion detection algorithms are used in networking devices, most of the IDS attacks are introduced in common networking devices such as router, switches, networking tapes etc. Researchers found various algorithms for detection of intruders in the network. At last, we arrives Machine Learning algorithms for detection of intruders in the network. Machine Learning approaches are rapidly emerging in various extents nowadays, But most of the algorithms results in the sarcastic manner due to its redundancies. In this paper, we surveyed huge number of existing systems regarding IDS and its impact in the network, the future of IDS is with the mixture of Machine Learning algorithms and its results in the detection of the intrusions with high accuracy.

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I. Introduction:

An Intrusion Detection System (IDS) is a device or software for the classifying and detect the cyber-attacks in network and host level to generating report by an administrator or centrally used security management systems (SIEM). This system merges the outputs from different sources, and uses Alarm Filtering (AL) technique to differentiate the malicious activity from False Alarms. Intrusion Detection System (IDS) technique is used in single computer to large computers network to detect the intrusion. The most commonly used IDS techniques are Network Intrusion Detection Systems (NIDS) and Host-based Intrusion Detection Systems (HIDS). In NIDS systems used by collecting information from various networking devices such as router, bridges, switches and networking taps. This will analyzed in order to identify attacks and most possible threats covered with in network traffic. IDS system uses varies system log to monitor the system activates which helps to detect attacks.

NIDS inspects the packets content in network traffic. For an HIDS systems gather information from local host computer which includes sensor logs, software logs, system logs, disk resources, and user account details for each system. Now a days many organization's uses hybrid of both NIDS and HIDS. Analysis of network traffic is done by misuse detection, anomaly detection, and state-full protocol analysis. Misuse detection uses signatures to detect the attacks, it collects human digital signatures and updated to the databases regularly to detect the attacks. Anomaly detection uses the heuristic method to detect the malicious attacks, but most of the times it gives the false results. In this case organization uses both misuse and anomaly method to detect the malicious attack effectively. State full analysis is most effective method comparatively other two methods. This method monitoring application layer, transport layer, and network layer. It uses predefined stakeholder specifications that stores in data base, to detect the unauthorized access/deviations in the

network protocols. By mapping the actual and predefined specifications to detected it. In recent days lot of challenges arise by malicious attacker are continuously changing their techniques to cracks the networks. It fined very large amount of scalable solutions.

II. Related works:

In NIDS uses the mathematics measures or computed thresholds it choices like packet length, bury purpose, flow size and totally different network traffic parameters are used in effective technique with during a} very special time window. but this model suffered from high rate of false positive and false negative. If the system as high negative false rate it's aiming to finishes up in fails within the detection within the network. thus it produces inefficient solutions to the systems. Throughout various machines learning primarily based solutions are found in literatures, it's aiming to be applicable to industrial systems is in early stages. the prevailing technic solutions are supported high false positive and high false negative rate it finishes up in system causes the high method worth. This draw back happens thanks to machine learning classifiers learn simple data sets from TCP/IP choices domestically. There are several hidden layer at intervals the TCP/IP protocol by learning techniques. a bit planned by Mukkamala et al. [1] compared the performance of SVM and ANN on the KDD CUP 9ty nine dataset. The results showed that the detection results of SVM are on top of those of ANN. In [2], SVM, naïve Bayes, provision regression, decision tree (DT), and classification and regression tree (CART) approaches were compared in terms of intrusion detection classification by mistreatment the KDD CUP 9ty nine dataset. The results showed that SVM has distinct choices.

Ashfaq et al.[3] presented an placement methodology by victimization in the indistinctness approach supported semi-supervised learning for intrusion detection. This methodology uses a

neural network with random weights and plays a very important role within the recognition rate of NIDS as a result of it decreases the process value. The model was evaluated on the NSL-KDD dataset however the performance of the model was studied on solely the binary classification task. Wang et al. [4] discussed an intrusion detection system known as hierarchic spatial-temporal feature-based intrusion detection system (HAST-IDS) network traffic is used DL (Deep Learning) Method such as Convolution Neural Network for erudition of low level temporal Method, LSTM networks (long remembering networks) for erudition of sophisticated temporal options. They used the quality DRAPA and ISCX2012 dataset to judge the performance of their planned system. Their model is computationally overpriced compared to our approach as a result of they used 2 stages for feature learning. Shamshirb and et al. [5] planned a hybrid clump methodology particularly Density-based Fuzzy Imperialist Competitive clump rule (D-FICCA) that may be a modification from density-based algorithm and symbolic logic to reinforce the accuracy of malicious detection like dos attacks. Then, Shamshirband et al. [6] planned Cooperative fuzzy artificial system (Co-FAIS) to mitigate the dos attacks. D-FICCA and Co-FAIS have the benefits to predict the presence of dos attack and equipped with the counter-defence mechanism for wireless device network. However, energy and memory consumption of D-FICCA and Co-FAIS in strained device like WSN node don't seem to be provided by the authors. W. Hu, W. Hu, and S. Maybank et al [7] planned Associate in Nursing ID rule victimisation the AdaBoost technique that wont to call stumps as weak classifier. this method performed better than different printed results with lower false rate with the next detection rate. This rule is computationally quicker than others. However, the rule is didn't adopt the incremented learning approach.

L. Ertöz, M. Steinbach, and V. Kumar et al [8] planned the recital of the shared nearest neighbour (SNN) primarily based ID model it completely was reported as a result of the most effective rule with high detection rate. It reduces the datasets they were able to report that SNN performed well as compared to the K-means for U2R attack. However, the system did not show the full dataset testing report. N. B. Amor, S. Benferhat, and Z. Elouedi et al [9] planned the rule mistreatment the Bayesian networks for ID. it completely was explored mistreatment Native Bayesian Networks with root node to represents a class of a affiliation and leaf nodes to expressed choices of a affiliation. Later, A. Valdes and K. Skinner et al [10] planned the careful experimental analysis with the Native Bayesian network to ID. They show the performance of Bayesian Networks equally any as typically even beyond U2R. D.-Y. Yeung and C. Chow, et al [11] planned a non-parametric density estimation methodology supported parzen- window estimators with Gaussian kernels and distribution. it's supported ensemble of decision trees. W. Li, et al [12] planned the Generic rule primarily based NIDS was facilities to the model for temporal and spatial- information to identify the difficult abnormal behaviour. C. Koliás, G. Kambourakis, and M. Maragoudakis et al [13] planned the model as Swarm intelligence techniques for IDS mistreatment insect colony optimization and Colony clump and particle swarm optimisation of systems. Justin Lee, Stuart Moskovics, film producer Arivudainambi et al [14] surveyed on IDS analysis procedure. They discusses regarding a pair of major problems in IDS square measure and rule-based behaviour analysis.

Attackers view:

Most of the time, intrusions are initiated by the attacker can attempt to gain access to a computer remotely via the internet and make services remotely unavailable. Generally an attacker can attacked the system by five ways they

are recognition, exploitation, reinforcement, consolidation, and pillage. It is very difficult to classify the normal system and attacked system. During the recognition phase the attacker can collect the information related to host and services, as well as they collection information about the operating system and application that are running in the system. During the exploitation they utilizes that particular information to attacks the target system. An attacker can do following activities they stolen the password of the system by various attacks such as brute force, dictionary attacks and SQL injection. After an illegal forced entry to a system, an attacker install supplementary tools and services to take privileges gained during reinforcement phase. They try to gain full access to the system in the consolidation phase. They possibly spread the malicious code to theft the data, CPU's time and impersonation during pillage phase. Since the networking devices are made by humans they have bug in the both hardware and software components in the networks mean by vulnerability. All the information technology security depends on the scalability, integrality, and confidentiality of computer networks.

Attackers and their signatures:

Attacker could leave there footprints by erroneously. that may be a signature identity of the network admin to spot, World Health Organization access that network. Some attacks are known by the signatures they're,

- Packet with associate degree nonlegal communications protocol flag combination, this technique are often known by examination the flags set during a communications protocol header by identified sensible or dangerous flag combos.
- Email containing a virulent effect, the IDS compare the subjective of mail and subject keep company with the virus mail, or it

will checks the attachment get in that specific mail.

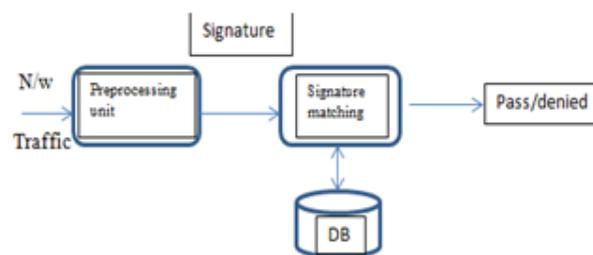
- DNS buffer over flow contains the payload of question, by passing the DNS fields and compare the length of every of the fields, it will makes an attempt to perform a buffer over flow victimization DNS fields.
- If admin desires to search out the wrongdoer physical address victimization reversed information processing address mechanism to find the attackers address in supply address field in associate degree IP header.

Most of the existing IDS are man-made it have some vulnerability in the systems. The man-made analyst is very difficult to detect the intrusion and non-intrusion network traffic. These inventions are based on common requirements like as create, coding, testing, and deploy the signature on the analyst data sets. It may take several hours to identify the intrusions which may leads to the inflexibility of the networks.

Signature Based Detection (SBD):

This method is used to comparing the observed signatures with signatures on the Data-Base (DB). That DB consists of known attack signature list. Through the feature extraction from any signature pattern the system monitored the packet and its environment. Signature matches on the database it will flag as a violation of the security policy. This method is little bit expensive in computation and preprocessing, so it does not monitor the every activity in the network traffic. Alternatively it only searches for known signatures in the database or file. SBD method also analyzes the system calls for the known threat payload. This method effective against the known attacks or violations but it is in-effective to detect the latest attacks until it is updated with latest signatures. This system can easily bypass by the attacker who knows the modification and selected

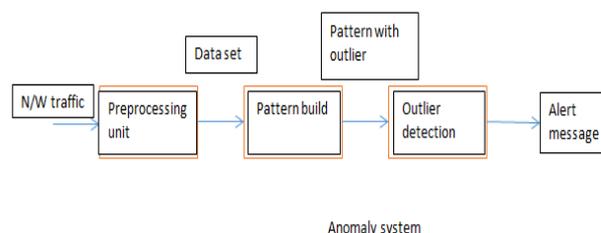
system which is not updated with latest signatures that detects the modifications.



Signature Based Architecture

Anomaly Based Detection:

This detection also called as Behavior – Based Detection (BBD) is trace as network IDS behavior model systems. It monitors the network traffic if any deviation from the normal behavior it raises an alarm to admin. Anomaly detection is used to detect a recent attack or recent potential attacks. It based on the principle of distance measured to construct profiles representing normal usage and then comparing with the latest/current behavior of the data's to find out a mismatch. Anomaly based detection method detects any traffic it can be unusual or new, and is good to identify pros and cons towards network hardware. These systems are easily detects the web-anomaly, port- anomaly to misfired Attacks, where the URL is mistyped. If any misclassifications it generate a lot of False Positive Alarms.



Anomaly Based Detection Architecture

According to the lecture survey lot of anomaly detection algorithms are used to prevent the systems from the attacker but most of the attacks are performed. Anomaly algorithms are produces result in high false positive ratios. So, that we move on to the machine leaning with the anomaly

detection in this methodology researchers uses the various algorithms to detect the attacks by trained datasets. Commonly known methods are generic algorithms, neural-networks; single learning techniques and support vector machines etc. in other hand combined techniques like hybrid detection algorithms are used.

Machine learning:

Machine learning technology for intrusion detecting systems used in recent decade for machine generated detecting systems. There are different types of malware data sets available in internet. Cyber security analyst community analysis those detailed and generated reports for various attacks and solve the solutions to it. This IDS the most changes occurs in the data set collected which contains many samples of intrusion techniques such as brute force attack, denial of service within the network. Most of the machine learning algorithms used to detect the intrusion in the networks. The common issues in existing solutions are high false positive rates in wide area of attacks, existing attacking solutions are based on single data set only but now huge amount of network traffics are occurs, existing solutions could not effectively detects the attacks. Commonly known machine learning IDs techniques are neural networks, Bayesian-networks, fuzzy logic and decision tree etc.

Neural networks:

The AI neural networks consists a set of processing elements it has highly interconnected and transforming input to desired output of the system. This neural network is based on the human biological nervous systems, such as brain and processing information and its works. For intrusion detection approach neural network uses multilayer perception (MLP) is widely used. This method is based on IDS deliberate to classify the normal and attacks pattern and various attacks. Neural networks have huge number of training set for computation facilities.

Decision Tree:

Decision Tree algorithm is used for classification. Here the datasets is learned and modeled, wherever a new data sets are classified, this classification is based on previous dataset decisions. This algorithm is also used for intrusion detection systems. The learned models are built in the system which classifies the attacks types and predicts the future data belongs to model built. The strength in this algorithm is it can process the massive data flow can identified across the computer networks. It uses generic algorithm to detect the IDS effectively. The generic algorithm primarily follows the given steps.

- The Intrusion Detection System (IDS) collects the knowledge concerning the traffic passing through a selected network.
- Then applies Genetic Algorithms that is trained with the classification rules learned from the knowledge gathered from the network analysis done by the IDS.
- To classify the incoming traffic by set of rules within the abnormal or traditional supported their pattern.
- GA biological process rule was with success employed in completely different IDS. It came back spectacular result; the simplest fitness worth was terribly closely to the perfect fitness value. it's a randomisation search technique typically used for improvement downside. This technique is with success generating a model with the required characteristics of high true detection rate and low false positive rate for IDS. And it used with success in Intrusion Detection System to tell apart the conventional action and also the intruded actions, and clump Gas could be a promising technique for the detection of malicious intrusions into pc systems.[14]

In these generic algorithms used to detect the intrusion detection in the networks by an efficient way.

Categories Of Attack	Attack name	Number Of Instances
DOS	SMURF	2807886
	NEPTUNE	1072017
	Back	2203
	POD	264
	Teardrop	979
U2R	Buffer Overflow	30
	Load Module	9
	PERL	3
	Rootkit	10
R2L	FTP Write	8
	Guess Password	53
	IMAP	12
	MultiHop	7
	PHF	4
	SPY	2
	Warez client	1020
	Warez Master	20
PROBE	IPSWEEP	12481
	NMAP	2316
	PORTSWEEP	10413
	SATAN	15892
normal		972781

Types of Attacks and their Occasions

In the recent decades a lot of Intrusion Detection attacks are performed in various categories. The most of the ID attacks are based on DOS, U2R, R2L, PROBE, and Normal attacks. These attacks are performed in various sub categories names they are mention in above table

Table VI. Average Accuracy Rate

Machine Learning Classifiers	Correctly classified Instances	Incorrectly classified Instances	Accuracy Rate
j48	55865	4135	93.10%
Random Forest	56265	3735	93.77%
Random tree	55345	5655	90.57%
Decision table	55464	4536	92.44%
MLP	55141	4859	91.90%
Naive Bayes	54741	5259	91.23%
Bayes Network	54439	5561	90.73%

Machine Learning Classifier and their Accuracy Rate

From the above table we understand that the machine learning classifiers are produces the

accuracy rate in IDS in very efficient manner. From these the machine learning algorithms are

used to the Intrusion Detection with high positive detection rate and reduce the false rate. Its accuracy is more accurate compare to the Intrusion Detection generic technics.

III. Conclusion:

In real world, network and its Security is viable to attackers due to enormous number of attacks arising day by day. In this paper, we arrived with conclusion that existing system for security and network has a lot of vulnerabilities. It is easy for attackers to find out those vulnerabilities of a Network. So, the algorithm has to develop with high accuracy and speed for detecting the intruders over the network, Machine Learning algorithm are developed rapidly, these algorithms for IDS results the high false rate in the previous techniques, here we minimizing the fault ratio and increase the performance of the system, so the Intruders Detection System has to develop with machine learning algorithm in near future. Different techniques are used to improve the performance of the system. In the immediate future risk is compromising the data or servers. The future for IDS is depends only on the machine learning concepts.

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