

AN ANALYTICAL STUDY FOR IDS SYSTEM BY KF MODEL USING IOT

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ABSTRACT

With the tremendous development of computer networks and the immense increment in the quantity of applications that depend on it, network security is increasing expanding significance. Besides, all computer systems experience the ill effects of security vulnerabilities which are both technically troublesome and economically expensive to be tackled by the manufacturers. Thusly, the part of Intrusion Detection Systems (IDSs), as unique reason gadgets to recognize anomalies and assaults in a network, is winding up more essential.

Customarily, intrusion detection techniques are grouped into two classes: abuse (signature-based) detection and inconsistency detection. Be that as it may, a few specialists have as of late proposed the possibility of hybrid detection to harvest the benefit of abuse detection by having a high detection rate on referred to intrusions and also the capacity of abnormality detectors in distinguishing shiny new assaults. In spite of the characteristic capability of hybrid detection, there are as yet two imperative issues that exceptionally influence the execution of these hybrid systems. Initially, oddity based techniques can't accomplish a remarkable execution without a far reaching named and progressive preparing set with all unique assault sorts, which is exorbitant and tedious to make if certainly feasible. Second, productive and compelling fusion of a few detection innovations turns into a major test for building an operational hybrid intrusion detection system.

Concerning the previously mentioned deficiencies, in this theory, we present a network-based intrusion detection system to perceive noxious network exercises and report them to the system administrator.

INTRODUCTION

The field of intrusion detection has gotten expanding consideration as of late. One explanation behind this is the hazardous development of the Internet and the vast number of networked systems that exist in a wide range of associations. The expansion in the quantity of networked machines has prompt an expansion in unapproved movement, from outer attackers, as well as from inward attackers, for example, disappointed representatives and individuals manhandling their benefits for individual pick up. Security is a major issue for all networks in the present venture environment. Hackers and intruders have made numerous fruitful endeavors to cut down prominent company networks and web services. Numerous methods have been produced to secure the network framework and communication over the Internet, among them the utilization of firewalls, encryption, and virtual private networks. Intrusion detection is a moderately new expansion to such techniques. Intrusion detection methods began showing up over the most recent couple of years. Utilizing intrusion detection methods, you can gather and utilize data from known sorts of assaults and see whether somebody is attempting to

assault your network or specific hosts. The data gathered along these lines can be utilized to solidify your network security, and also for lawful purposes. Both business and open source items are currently accessible for this reason. Numerous helplessness evaluation devices are additionally accessible in the market that can be utilized to survey diverse sorts of security openings display in your network. Over the most recent couple of years, various intrusion detection systems have been produced both in the business and scholarly areas. These systems utilize different ways to deal with identify unapproved action and have given us some knowledge into the issues that still must be illuminated before we can have intrusion detection systems that are valuable and solid underway settings for recognizing an extensive variety of intrusions. With the exponential development of the Internet and networked computers, digital wrongdoing has turned out to be a standout amongst the most vital issues in the computer world. Online charge card misrepresentation, traded off computer servers and other protection enormities have made a billow of doubt among online clients. As per CSI/FBI Computer Crime

and Security Survey, the aggregate income misfortune in industry because of computer network intrusion was figured as \$455,848,000, up from \$35 million revealed. These numbers legitimize the expansion in inquire about enthusiasm for computer security.

LITERATURE REVIEW

S. F. Owens and R. R. Levary have expressed that intruder detection systems have been regularly constructed utilizing expert system innovation. In any case, Intrusion Detection System (IDS) researchers have been one-sided in constructing systems that are hard to handle need sagacious user interfaces and are badly arranged to use in real-life circumstances.

Alok Sharma has concentrated on the utilization of content processing techniques on the system call arrangements for intrusion detection. Host-based intrusions have been identified by introducing a kernel based similarity measure. Processes have been grouped either as normal or abnormal utilizing the k-nearest neighbor (kNN) classifier.

Shi-Jinn Horng has used a combination of hierarchical clustering algorithm, easy feature selection method, and SVM

technique in their proposed SVM-based intrusion detection system. Fewer, abstracted, and higher-qualified training instances that are derived from the KDD Cup 1999 training set has been given to the SVM by the hierarchical clustering algorithm.

B. Shanmugam and Norbik Bashah Idris have proposed an advanced fluffy and data mining methods based hybrid model to discover both abuse and anomaly attacks. Their goal was to decrease the quantity of data kept for processing and likewise to improve the detection rate of the current IDS utilizing attribute choice process and data mining strategy respectively. O. A. Adebayo has presented a strategy those utilizations Fuzzy-Bayesian to identify real-time network anomaly attack for discovering noxious movement against computer network. They have set up the adequacy of the technique by describing the framework. The overall performance of the intrusion detection system (IDS) based on Bayes has been improved by a mix of fluffy with Bayesian classifier. Moreover, by the experiment carried out on KDD 1999 IDS data set, the practicability of the technique has been

verified.

Abadeh, M.S. and Habibi, J. has proposed a strategy to create fluffy grouping rules for intrusion detection use in computer networks. The strategy for fluffy rule base system configuration has been based on the iterative rule learning approach (IRL). Utilizing the evolutionary algorithm to advance one fluffy classifier rule at a time, the fluffy rule base has been created in an incremental manner.

Arman Tajbakhsh has presented a data mining procedure based framework for constructing IDS. In the framework, Association Based Classification (ABC) has been utilized by the arrangement motor which is in actuality the central part of the IDS. Fluffy affiliation rules have been utilized by the proposed arrangement to construct classifiers.

Zhenwei Yu has presented an automatically tuning intrusion detection system (ATIDS). According to the criticism provided by the system operator, when false predictions are recognized, the proposed system automatically tunes the detection model on-the-fly.

RESEARCH METHODOLOGY

This is the data set utilized for The Third International Knowledge Discovery and Data Mining Tools Competition, which was held in conjunction with KDD-99. The Fifth International Conference on Knowledge Discovery and Data Mining. The opposition assignment was to fabricate a network intrusion finder, a prescient model fit for recognizing "bad" associations, called intrusions or assaults, and "good" typical associations. This database contains a standard set of data to be inspected, which incorporates a wide assortment of intrusions reproduced in a military network environment.

In 1998, DARPA intrusion detection assessment program, a reenacted environment was set up to gain crude TCP/IP dump data for a neighborhood (LAN) by the MIT Lincoln Lab to think about the execution of different intrusion detection methods. It was worked like a genuine environment, however being impacted with numerous intrusion assaults and got much consideration in the exploration group of versatile intrusion detection. In KDD99 dataset, every illustration speaks to quality estimations of a

class in the network data stream, and each class is named either ordinary or assault.

The classes in KDD99 data set can be arranged into 5 primary classes (one typical class and four principle intrusion classes: test, DOS, U2R, and R2L).

1. Normal associations are created by recreated day by day client conduct, for example, downloading files, going by web pages.
2. Denial of Service (DoS) assault causes the processing force or memory of a casualty machine excessively occupied or too full, making it impossible to deal with genuine solicitations. DoS assaults are characterized in light of the services that an assailant renders inaccessible to authentic clients like apache2, arrive, mail bomb, back, and so on.
3. Remote to Local (R2L) is an assault that a remote client obtains entrance of a nearby client/account by sending packets to a machine over a network communication, which incorporate send-letters, and Xlock.
4. User to Root (U2R) is an assault that an interloper starts with the entrance of a typical client record and afterward turns

into a root-client by abusing different vulnerabilities of the system. Most normal adventures of U2R assaults are consistent cushion overflows, stack module, Fd-configuration, and Ffb-config.

5. Probing (Probe) is an assault that sweeps a network to accumulate information or find known vulnerabilities. A gatecrasher with a guide of machines and services that are accessible on a network can utilize the information to search for abuses.

In 1999, the first TCP dump files were preprocessed for usage in the Intrusion Detection System benchmark of the International Knowledge Discovery and Data Mining Tools Competition. To do as such, bundle information in the TCP dump record is condensed into associations. In particular, "an association is a succession of TCP packets beginning and closure at some very much characterized circumstances, between which data flows from a source IP deliver to an objective IP address under some all around characterized convention". This procedure is finished utilizing the Bro intrusion detection system, bringing about 41 highlights for every association.

Components are assembled into four classifications:

- **Basic Features:** Basic features can be gotten from bundle headers without examining the payload. Basic features are the initial six features recorded in Table 5.2
- **Content Features:** Domain information is utilized to survey the payload of the first TCP packets. This incorporates features, for example, the quantity of fizzled login endeavors
- **Time-based Traffic Features:** These features are intended to catch properties that develop over a 2 second fleeting window. One case of such an element would be the quantity of associations with a similar host over the 2 second interim
- **Host-based Traffic Features:** Utilize a verifiable window assessed over the quantity of associations for this

situation 100 rather than time. Host based features are in this manner intended to survey assaults, which traverse interims longer than 2 seconds.

ANALYSIS

The data set used to play out the analysis was taken from KDD Cup '99, which is generally acknowledged as a benchmark dataset. "10% of KDD Cup'99" and "revised (test)" from KDD Cup '99 data set was assessed execution of our strategy as preparing and testing data sets to identify intrusion.

A "10% of KDD Cup'99" circulation records as preparing dataset by class sort is condensed in Table 5.4. In the interim Table 5.5 demonstrates the testing dataset information acquired from "Amended (Test)". The conduct of data for intrusion detection system can be ordered as in Table 5.6.

Class	No. of Samples	Sample percentage
Normal	97277	19.69
Probe	4107	0.83
DoS	391458	79.24
U2R	52	0.01

R2L	1126	0.23
Total	494020	100.00

Table 1: Sample distribution of the training data set

Class	No. of Samples	Sample percentage
Normal	60593	19.40
Probe	4166	1.33
DoS	231455	74.40
U2R	88	0.028
R2L	17727	4.73
Total	311029	100.00

Table 2: Behavior of data

Intrusion detection system require high detection rate and low false alarm rate, in this manner the execution of a Network Intrusion Detection System can more often than not be assessed as far as exactness, detection rate and false alarm as underneath:

$$Accuracy = \frac{TP + TN}{TP + TN + FP + FN}$$

$$Detection\ Rate = \frac{TP}{TP + FP}$$

$$False\ Alarm = \frac{FP}{FP + TN}$$

Where,

FN is False Negative,

TN is True Negative,

TP is True Positive, and

FP is False Positive

A progression of trials was directed utilizing the created IDS arrangement approach with the benchmark dataset, KDD-Cup'99. All data was standardized and a few features have been changed before the execution to acquire a superior yield. Cross-approval is a

standout amongst the most ordinarily utilized methods. In 10 cross-approvals the entire dataset will be isolated into 10 subsets, which 9 subsets consider in the training subsets and the rest as the testing subset. The results are executed by five classification classes Normal, Probe, DoS, U2R and R2L.

RESULTS

As specified before, KDD Cup '99 dataset is utilized to assess the proposed way to deal with contrast and Naïve Bayes classifier. There are two parts of data set utilized,

which are training data and testing data. The test data contain an unforeseen assault which has not been canvassed in the training set.

Table 5.7 and Table 5.8 demonstrate the confusion network of innocent bayes classifier and Decision Tree based created IDS. A "Confusion Matrix" is sometimes used to speak to the consequence of training and testing, as appeared in Table 5.7 and Table 5.8. The Advantage of utilizing this lattice is that it discloses to us what number of got misclassified as well as what misclassifications happened.

Actual	Predicted Normal	Predicted DoS	Predicted Probe	Predicted U2R	Predicted R2L	Accuracy (%)
Normal	8909	8	138	570	102	91.6
DoS	444	36921	16	1757	8	94.3
Probe	0	0	410	0	1	99.8
U2R	0	0	0	4	1	80.0
R2L	27	0	3	9	74	65.5

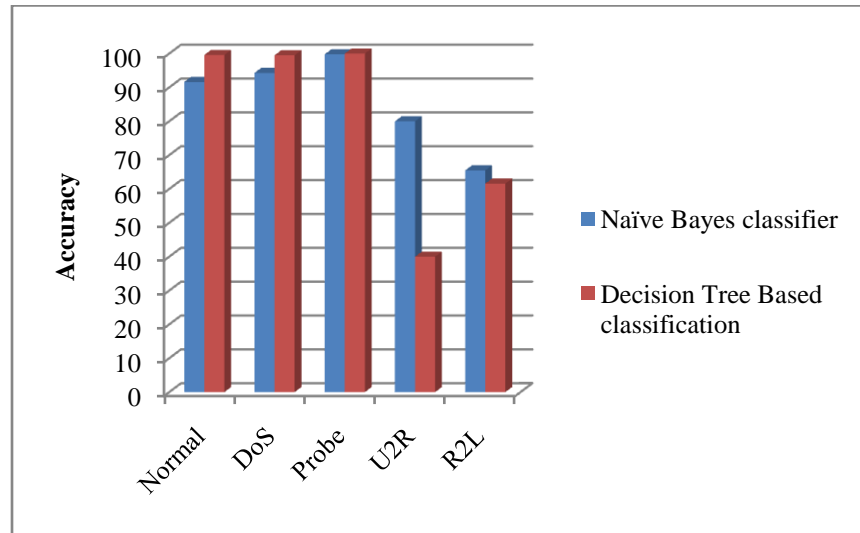


Figure 1: Accuracy comparison graph by using training data set

A movement of trials was coordinated using the made IDS course of action approach with the benchmark dataset, KDD-Cup'99. All data was institutionalized and a couple of features have been changed before the execution to gain an unrivaled yield. Cross-endorsement is a champion among the most normally used methods. In 10 cross-endorsements the whole dataset will be separated into 10 subsets, which 9 subsets consider in the training subsets and the rest as the testing subset. The results are executed by five order classes Normal, Probe, DoS, U2R and R2L.

Results analysis for Training Dataset

Table 5.7 and Table 5.8 exhibit the confusion network of guiltless bayes classifier and Decision Tree based made IDS. A "Confusion Matrix" is sometimes used to address the outcome of training and testing, as showed up in Table 5.7 and Table 5.8. The Advantage of using this cross section is that it uncovers to us what number of got mis-delegated well as what mis-orders happened..

Actual	Predicted Normal	Predicted DoS	Predicted Probe	Predicted U2R	Predicted R2L	Accuracy (%)
Normal	7875	14	131	1664	43	81.0

DoS	6431	32298	417	0	0	82.5
Probe	6	12	393	0	0	95.6
U2R	1	0	0	4	0	80.0
R2L	10	00	1	0	102	90.3

CONCLUSION

The recommended approach called Decision Tree Based characterization is assessed and contrasted and the single Naïve Bayes classifier utilizing KDD Cup '99 data set. The exploratory outcomes demonstrate that the k Decision Tree Based characterization approach accomplishes better exactness and detection rates while lessening the false alert by identifying novel intrusions precisely. The execution of Naïve Bayes classifier has been enhanced by applying Decision Tree Based order. Notwithstanding, Decision Tree Based characterization has restriction to identify intrusions that are fundamentally the same as with each other, for example, U2R and R2L.

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