



A REVIEW ON KDD CUP99 AND NSL-KDD DATASET

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Abstract: Continues use of network services for information and resource sharing makes our work easier. But sometime the extensive use of network services leads many problems in the form of attacks or intrusions which demolish not only the privacy but also the integrity and accessibility of data. Detection of attacks or intrusions on the network is a serious issue of concern for the researchers. Intrusion Detection System solves the purpose of detecting intrusion on the network. Huge amount of data is required to simulate the powerful Intrusion Detection System (IDS) model as well as to train and testing the model. This paper, presents the review of datasets DARPA, KDDCup99 and NSL_KDD which are most widely used by researchers to detect the intrusion in computer network.

Keywords: IDS, Dataset, DARPA, NSL_KDD, KDDCup99.

INTRODUCTION

Now days, the use of network is growing due to the increased use of handheld devices. So, network security is the major issue. Our network suffers from various types of attacks like viruses, Trojan horse, worms. To identify and stop these attacks, a security management system is required. Confidentiality, Integrity and availability of data are the major objectives of security. An Intrusion Detection System serves this purpose by automatically alert the administrator when someone trying to violets the security policies. The role of intrusion detection system is to assemble the information from the network. Then after supervising and investigating this information, it separates them into normal & malicious behaviour and brings this result to system administrator [2].

Encryption, firewalls, virtual private network etc. are the conventional approaches which were used in early days. But they were not able to protect the network completely. Thus to increase the network security, an Intrusion Detection System is introduced. It is divided into two main categories as Signature and Anomaly based . Another name of signature based IDS is misuse based IDS. It identifies only the familiar attacks. But anomaly based IDS can identify known as well as novel attacks.

To figure out the conduct of Network Intrusion Detection System, various datasets are available. To provide the security to computer network, many researchers have suggested three most widely used datasets like DARPA 98/99, KDD99 and NSL- KDD. DARPA is the first dataset for the evaluation of intrusion detection system and was attempted in MIT Lincoln Laboratory in 1998. KDD CUP99 is the subset of DARPA98 dataset. It has 41 features. NSL-KDD dataset is derived from KDDCup99 by

removing the redundant and duplicate records from training and testing datasets respectively so it is the revised version of the original KDDCup99 dataset. Each dataset has its advantages and shortcomings. It is very challenging to select a suitable dataset itself. Due to the increased use of network, the behaviour and the pattern changes and dependency on a particular dataset is not trust-worthy. So, there is need to update the dataset periodically.

In this paper the review of DARPA, KDD Cup99 and NSL_KDD datasets are made using various attributes.

DATASETS

KDDCup99 dataset (Knowledge Discovery in Databases):

The KDDCup99 is the mostly and commonly used dataset for the identification of intrusion in computer network. Simulation of US Air Force LAN was done in order to get the subset of DARPA 1998 dataset by inducing different types of attacks. A nine week of TCP dump data was used for this purpose at MIT Lincoln Laboratory. KDDCup dataset contains about 4,900,000 single instances which are described by 41 features [1]. They are classified as either normal or an intrusion.

Table1. Features in the KDD Cup99 Dataset

Sr. No.	Characteristics	Description
1	Kind of Protocol	Description of the protocol used
2	Time intervals	Time interval of the connection

3	Src_bytes	number of bytes to be transferred from source
4	Service	Service on the destination
5	Urgent	statistics of important packets
6	Flag	Gives the description about the status whether normal or error
7	Dst_bytes	number of bytes to be transferred from destination
8	Wrong_fragmant	Number of wrong segments
9	Land	When IP address of sending end, receiving end and port number are equal then set 1 otherwise 0.
10	Num_file_creation	Total operation to create a file
11	Root_shell	Whenever root account is running then set 1 otherwise 0
13	Hot	statistics of hot indicators
14	Su_attempted	Whenever su command is used then set 1 otherwise 0
15	Logged_in	If login successfully then set to 1 otherwise 0
16	Num_root	Statistics of the performed operations as a root
11	Num_failed_logins	Total of failed login attempt
13	Num_compromised	Total of Compromised situations
14	Num_access_files	Total actions required to access the control files
15	Num_outbound_cmds	In the ftp period, number of outgoing commands
18	Num_shell	Total shell stimulate
21	Is_hot_login	Whenever signed through hot list set to one else zero

22	Count	In past two seconds, the total of connection to the similar destination node as the ongoing connection
23	Is_guest_login	If signed through guest then set 1 otherwise 0
24	Serror_rate	% of SYN fault in connections
25	Srv_count	Describe the total connections in past two second to the similar service (port number) as the ongoing Connection
27	Srv_error_rate	% of REJ fault in connections
26	Srv_error_rate	% of SYN fault in connections
27	Error_rate	% of REJ fault in connections
28	Diff_srv_rate	% of variation in connections
29	Same_srv_rate	% of connections to the similar service
32	Dst_host	Count number of connections to the same destination
31	Srv_diff_host_rate	% of connections to different hosts
33	Dst_host_srv_count	Count the connection with same destination and also use the same service
34	Dst_host_same_srv_rate	% of connections to the same service and destination
35	Dst_host_srv_serror_rate	% of connections to a host as well as specified service with an S0 error
36	Dst_host_same_src_port_rate	% of connections on the same source port
37	Dst_host_srv_diff_host_rate	% of connections on different destination
38	Dst_host_diff_srv_rate	% of connections to various services
39	Dst_host_serror_rate	% of S0 fault in a host

40	Dst_host_srv_error_rate	Connections in % that triggered the flag (4) REJ
41	Dst_host_error_rate	% of connections that triggered the flag (4) REJ

Features of KDD dataset are classified into four different classes as:

Basic: This class consist of all the features of TCP connection.

Content: This class consist of all the features given by domain knowledge within a connection.

Traffic: this class consist of features which are computed within a time frame of two seconds.

Host: This class consist of features which last for more than two seconds.

All the attack are divided into one of the following categories[3]:

Denial of Service Attack (DoS): In this attack our system resources knowingly occupied by some unnecessary or unwanted processes in order to make the server too busy to handle the other important requests which result in rejection of legitimate request.

User to Root Attack (U2R): In this type of attack, the intruder tries to gain the access of an authorized user account in the system and exploit some vulnerability to gain super user privilege.

Remote to Local Attack (R2L): A person who doesn't have an account on a machine but yet sending packets to the same machine on a network to personify the legal user for gaining the local access to the machine.

Probing Attack: These attacks gather the network activity information for the supposed objective to bypass its security controls.

Advantages of KDDCup99:

KDDCup99 dataset has some improvement over DARPA 1998 dataset:

- Conversion the network traffic from TCP dump file into relational structure is not required.
- The dataset contains direct and derived features which are readily available.
- The memory and processing power is less required.
- To optimize accuracy and detection rate over KDD99 dataset many machine learning algorithms are used. Most frequent algorithms used with KDD99 are decision tree derivatives and support vector machines.

Problems of KDD99 Data Set:

- The synthesized data is not matching to real traffic of network data.
- Training and test sets are too large which make it very complex.
- Detection accuracy is very low.
- Cannot be detect dropped packets.

- Unreliable for building real NIDS.
- It has redundant and duplicate records
- Machine learning algorithm cannot be applied to R2L and U2R.
- Large gap between the number of instances of normal traffic and number of instances of attack.

NSL-KDD Dataset: This dataset designed by Tavallae et al. [5]. It is developed after the removal of redundant and duplicate records from training and test data of KDDCup . it contains only selected and necessary records from. There are total

37 attacks out of which 27 attacks are used by testing dataset and 23 attacks are used by training dataset for experiments[8]. The number of feature in NSL-KDD dataset has same as that of in KDDCup. This dataset contains 41 features and 5 attack classes. There is one normal class and other

4 are different types of attack. These different attacks are grouped into four categories: Probe attack, Denial of service attack(DoS), User to Root (U2R) and Remote to Local (R2L). The above dataset holds a binary class attribute as well as reasonable number of training and test instances [6]. This dataset is publically available for researchers.

Advantages of NSL-KDD:

- Removal of redundant records helps the classifier to produce unbiased result.
- Since not even a single record found identical in proposed test set; therefore, learner's performance is not biased by the methods having better detection rates on the frequent records.
- Detection rate is high as compared to KDD Cup.
- The record counts in the train and test sets are reduced. Therefore selection of a chunk of data is not required randomly and all the experiments can be done on the entire set. As well as it gives consistent result of different research work.

There are total 21 different types of attacks which are present in training dataset. While test dataset contain 16 additional attacks. Major attacks are categorised as Probe, DoS, U2R and R2L[7].

Table II. Categories of Attacks for Training And Testing Datasets

DOS	Probe	R2L	U2R
Back	Ipsweep	Spy	bufferoverflow
Land	Mscan	Warezcilent	Loadmodule
Mailbomb	Nmap	ftp_write	Perl
Neptune	PortswEEP	Guesspassword	Ps
Pod	Saint	Httpunnel	Rootkit

Processable	Satan	Imap	Snmptguess
smurf		Multihop	Sqllattack
Teardrop		Named	Worm
udpstorm		Phf	xterm
		Sendmail	
		Snmptgetatt ack	
		Warezmaste r	
		Xlock	
		Xsnoop	

CONCLUSION

In order to develop new tools and in the research area of IDS KDD Cup99 is most known dataset for the protection of computer network against malicious activities. This dataset also have many limitations like redundant and duplication of records, imbalance between normal traffic and number of attacks and many more listed above. The solution of this is NSL-KDD which has removed unnecessary and same records in both training and test sets. Continuous use of computer network and information system has become the vital source for large number of attacks. Now a days , in all over the world, many researcher are developing new datasets by taking the help from KDD Cup, NSL- KDD and DARPA datasets depending upon the issues in problem solving and purpose of IDS.

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