# RESEARCH ARTICLE <br> A SURVEY OF INTRUSION DETECTION TECHNIQUE USING VARIOUS TECHNIQUE OF MACHINE LEARNING 

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#### Abstract

Cyber's terrorism has recently been said to be the biggest threat to our modern society. Every day a new cyber-scare story makes the headlines. The national government recognizes the importance of cyber security, as several officials have made clear in the past few years. Cyber security is among the most serious economic and national security challenges we will face in the 21 st Century, we face a long-term challenge in cyberspace from foreign intelligence agencies and militaries, criminals, and others, and, struggle will wreak serious damage on the economic health and national security. For the prevention and detection of cyber terrorism intrusion detection system has been used. Intrusion detection system detects illegal behavior of network over data. In current research trend performance of intrusion detection system is important issue. Now various authors used machine learning and feature optimization technique for intrusion detection system. Machine learning technique is collection of all learning algorithm such as classification, clustering and regression. For the improvement of machine learning technique feature optimization techniques has been used. This paper presents review of intrusion detection techniques using machine learning and feature optimization process.


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## INTRODUCTION

The advancement of information communication network contributes to improvement of the quality of daily life of people, and now considered as fundamental social and economic infrastructure. However, the increase of incidents and threats against this infrastructure has turned out to be a serious problem. These days it is very significant to maintain a high level security to ensure protected and trust information message among different groups. But integrity of data over internet and any other network is always under threat of intrusions and misuses. So Intrusion Detection Systems (IDS) have become crucial components in computer and network security (Mohammad A. Faysel and Syed S. Haque, 2010). Improvement of intrusion detection technique is major concern of financial sector and social networking site for the use of common user.Computer security community has developed a variety of intrusion detection systems to prevent attacks on computer systems. Feature optimization and feature reduction is major challenges in current researcher trend in intrusion detection technique. Irrelevant and redundant attributes of intrusion detection data-set may lead to complex intrusion detection model as well as reduce detection accuracy.

[^0]The network based intrusion detection is called as mysterious attacks and this attack is analyzed on the basis of normal attack scenario (Jonatan Gomez and Dipankar Dasgupta, 2002). Despite all the applied mechanism for intrusion detection system, it does not provide the complete secured data. Therefore, intrusion detection is becoming an increasingly important technique that monitors network traffic and identifies network intrusions attacks to computer systems. A number of machine learning based approaches have been used for detecting abnormal threats. Machine learning refers to a group of techniques that develop the easiness for ambiguity, improbability, incomplete fact, and estimate to achieve toughness and low cost solution. The principle constituents of machine learning are Fuzzy Logic (FL) (Bapuji et al., 2012; Ajith Abraham et al., 2005), Artificial Neural Networks (ANNs), Probabilistic Reasoning (PR), and Genetic Algorithms (GAs). The Genetic Algorithm is used to detect the intrusions in networks (Li Liu et al., 2014; Ren Hui Gong et al., 2005). It considers both temporal and spatial information of network connections during the encoding of the problem using Genetic Algorithm. Data mining is an efficient method for intrusion detection, which can dig out the unknown knowledge and rules from a large number of network data or audit data from host. The Genetic Algorithm is more helpful for identification of network anomalous behaviors. The Rough Set Neural Network Algorithm is used to reduce a number of computer resources
required to detect an attack. The KDDCup'99 data-set is used to test the data and gives the better and robust result. The various feature reduction techniques such as Independent Component Analysis, Linear Discriminate Analysis and Principal Component Analysis used to reduce the computational intensity. KDD cup 99 data-set is used to reduce computation time and improves the accuracy of the systems. Section-I gives the introduction of the intrusion detection. Section-II gives the traffic feature of intrusion detection system. Problem formulations in intrusion detection have been reviewed in section-III. Section IV discusses comparative result evaluation. Finally, section-V is the conclusion and future scope of work.

## Traffic feature of network

The generation of network traffic is very large amount, processing of this traffic data is very difficult for firewall, intrusion detection system and other security analysis of tools. The generated traffic is not formatted, due to this reason the classification of traffic categories are very difficult. For the analysis of traffic data KDD mining tools are used and converted into connection and sequence data (Dewan Md. Farid et al., 2008). These sequence and connection data have 42 features on different categories such as basic feature data, content feature, time based feature and host traffic based feature.

1. Basic Features: - the basic feature of these categories gets information from packet header without information of payload. The content of these categories is 1 to 8 .
2. Content Features: - In this group new TCP packets analyzed with the help of area information. An example of this category is number of "hot" indicator.
3. Time-based Traffic Features: - for gathering these types of features a window of 2 second interval is defined. In this interval, some properties of packets are measured. For example number of connections to the same service as the current connection in the past two seconds.
4. Host-based Traffic Features: - In this category instead of a time based window, a number of connections are used for building the window. This category is designed so that attacks longer than 2 second can be detected.

The processing of feature and description of feature discuss in Table 1, 2 and 3 according to their description and data type.

Feature selection and feature reduction is an important data processing step prior to performing intrusion detection technique (Srilatha Chebrolu et al., 2005). Feature optimization and feature reduction process used some heuristic function such as genetic algorithm, particle of swarm optimization and neural network. In the all categories of feature some features play ideal role in connection stream in case of normal connection and abnormal connection. If reduces these feature it improves the performance of intrusion detection technique.

## Problem formulation

The environment in which the feature extraction and feature reduction is done is crucial section for intrusion detection. This means that the network traffic contains user confidential information. In general, only the header fields of the packets
can be checked but not the user data in the payload. Scalability is an issue with IDS.

Table 1. Basic features of individual TCP connections

| Feature name | Description | Type |
| :---: | :---: | :---: |
| hot | number of "hot" indicators | continuou <br> s |
| num_failed_logins | number of failed login attempts | continuou s |
| logged_in | 1 if successfully logged in; 0 otherwise | discrete |
| num_compromised | number of '"compromised' ' conditions | $\begin{aligned} & \text { continuou } \\ & \mathrm{s} \end{aligned}$ |
| root_shell | 1 if root shell is obtained; 0 otherwise | discrete |
| su_attempted | 1 if "su root" command attempted; 0 otherwise | discrete |
| num_root | number of "root" accesses | $\begin{aligned} & \text { continuou } \\ & \mathrm{s} \end{aligned}$ |
| num_file_creations | number of file creation operations | continuou $\mathrm{s}$ |
| num_shells | number of shell prompts | continuou s |
| num_access_files | number of | continuou |
|  | operations on access control files | s |
| $\begin{aligned} & \text { num_outbound_cmd } \\ & \mathrm{s} \end{aligned}$ | number of outbound commands in an ftp session | continuou <br> s |
| is_hot_login | 1 if the login belongs to the hot" list; 0 otherwise | discrete |
| is_guest_login | 1 if the login is a "guest"login; 0 otherwise | discrete |

Because of the huge amount of data flowing through the mobile operator's network, it is not an easy task to find out the right information needed for IDS. The problem is tofind an answer to the question: "What features need to be taken into account when calculating or analyzing whether the activity is malicious or not?" Based on prior research on IDS it is clear that either one of the techniques alone cannot detect everything but the combination of the both is the most promising approach. For example misuse detection can be used to filter known threats from the traffic to make it easier for the anomaly detection system to focus on the unknown. Even though IDS have been researched over 20 years, we still do not have an answer to the question of what features should be monitored. So far different kinds of methods and algorithms have been developed for anomaly detection but the focus has been on making them more efficient. Almost all of them are lacking the same information; what features are important for IDS, especially in telecommunications networks? For some reason information on the used features is not easily found
from IDS research publications. No matter what the reason is the result is the same; every researcher has to figure out by themselves which features should be used for the monitoring.

1. The pre-processing of KDDCUP99 takes more time.
2. The rate of false alarm generation is high.
3. Some data mining classifier are ambiguous situation for selection of base classifier
4. Entropy based intrusion detection system suffered by high false rate
5. The detection of dynamic feature evaluation.

## Comparative study of detection technique

In this section discussion of the comparative study of intrusion detection based on machine learning and feature optimization have been performed. This study focused on different methods of detection rate and finally demerits of the methods has been emphasized.

Table 2. Content features within a connection suggested by domain knowledge

| Feature name | Description | Type |
| :---: | :---: | :---: |
| count | number of comnections to the same host as the curreut connection in the past two seconds | continuous |
| serror_rate | \% of connections that have "SYN" errors | continuous |
| rerror_rate | \% of connections that have "REJ" errors | continuous |
| same_sry_rate | $\%$ of connections to the same service | continuous |
| diff_5rv_rate | $\%$ of connections to different services | continuous |
| 5rV_coumt | number of connections to the same service as the current connection in the past two seconds | continuous |
| $\begin{aligned} & \text { srv_serror_rat } \\ & \text { e } \end{aligned}$ | \% of connections that have "SYN" errors | continuous |
| $\begin{aligned} & \text { srv_rerror_rat } \\ & \text { e } \end{aligned}$ | $\%$ of connections that have "REJ" errors | continuous |
| $\begin{aligned} & \text { srv_diff_host_ } \\ & \text { rate } \end{aligned}$ | $\begin{aligned} & \text { \% of } \\ & \text { connections to } \end{aligned}$ | continuous |

Table 3. Traffic features computed using a two- second time window

| Feature name | Description | Type |
| :--- | :--- | :--- |
| count | number of connections <br> to the same host as the <br> current connection in <br> the past two seconds | continuous |
| serror_rate | \% of connections that <br> have "SYN" errors | continuous |
| rerror_rate | \% of connections that <br> have "REJ" errors | continuous |
| same_srv_rate | \% of comnections to the <br> same service | Continuou <br> 5 |
| diff_sry_rate | \% of comections to <br> different services | Continuou <br> s |
| srv_count | number of comections <br> to the same service as <br> the current connection <br> in the past two seconds | Continuou <br> 5 |
| srv_serror_rate | \% of connections that <br> have "SYN" errors | Continnou <br> 5 |
| srv_rerror_rate | \% of connections that <br> have "REJ" errors | Continnou <br> 5 |
| srv_diff_host_ <br> rate | \% of comnections to <br> different hosts | continuous |

Table 4. Comparative study of different intrusion detection techniques

| $\begin{gathered} \mathrm{SI} \\ \mathrm{No} \end{gathered}$ | MethodAp proach used | Rate of deduction | Demerits |
| :---: | :---: | :---: | :---: |
| 1. | Fuzzy Grenetic algorithm [1] | Avarage derection rate $97.00 \%=$ | Used only Limited number of data attribure |
| 2. | Neural nerwork classifier [2] | ```Approvim ate efficiency 97.50%=``` | Efficiency can be improve further with using featare Ieduction. |
| 3. | Leaming <br> Lamistar <br> Neural <br> Verworls [4] | Improved efficiency 98.009 $=$ | SOM which shows poor result for probe class. |
| 4. | Decision tree <br> classifier <br> with <br> feature <br> based GA <br> rechnicques <br> [5] | Improved classificati on rate | ```Decision tree not included the forest tree.``` |
| 5. | Feature retuction with FNTN and Blyes classifier [7] | Eliminate <br> non-useful <br> inñormatio <br> - based <br> fearure | Sometimes Computation cost may not be supported |
| 6. | Randiom forest classifier with SMOTE | $\begin{aligned} & \text { Build a } \\ & \text { model in } \\ & \text { Resstime } \end{aligned}$ | Nat aseful for Real time adaptive ID System |


| 7. | Generic Approach [9] | Improved model for misuse detection systems | Method support only some feature in $\operatorname{KDDCLP}$ |
| :---: | :---: | :---: | :---: |
| 8. | ```Data moining and newrai network approach [10]``` | Detection rate for $R B F$ is $9850 \%$. | Fesults not compared with any PCA. Aproach |
| 9. | Artribute selection and classificatio II $\quad$ Isethod [11] | Gives Accuracy is $99.00 \%$ | Model does not with Realy world |
| 10. | Unsupervis ed Neural network [13] | Accaracy $97.00 \%$ with ART and 95.00\% SOM nets | ART-2 offered a little lower detection rate performance than ART-1 |
| 11. | Genetic Algorithm [14] | Accuracy for training data set is $97.25 \%$ and testing ser is $95.00 \%$ | For the Inge training rate of datasers, it is neither efficient mor feasible |
| 12. | Fuzzy <br> Based <br> Divide <br> Conguer <br> Algorithm <br> [16] | Classificat ion rate is 99.96\% | Works only for selected features |
| 13. | Feanure <br> selection <br> and <br> Ensemble <br> method <br> [16] | Accuracy for Probe is $100 \%$ and for $R 2 L \quad$ is $99.97 \%$ | U2R gives <br> not berter <br> result, it's <br> suffered  |
| 14 | $\begin{aligned} & \text { FNNN-SVM } \\ & {[21]} \end{aligned}$ | Average accuracy p erforman ce is $97 \%$ | Very complex model used more time for evecurion |

## Conclusion and Future scope

In this paper study of intrusion detection technique using machine learning and feature optimization technique have been surveyed. In this study it has been observed that features of network data is very complex due to mixed categories. For the classification task feature selection and feature optimization re important techniques. Optimization of features and reduction of features will be carried out using neural network techniques, genetic algorithm and particle of swarm optimization techniques. Ithas also shown that merging a different classification technique also improves the detection of intrusion detection. All the methods used in this survey gives average $98 \%$ detection rate with false alarm generation. In future increasedetection rate will be increased approximately $100 \%$ and reduces the false alarm generation.

## REFERENCES

Abebe Tesfahun and D. Lalitha Bhaskari, 2013. "Intrusion Detection Using Random Forests Classifier With Smote And Feature Reduction" International Conference On

Cloud \& Ubiquitous Computing \& Emerging Technologies, Pp 127-133.
Ajith Abraham, Ravi Jain, Johnson Thomas and Sang Yong Han, 2005. "D-Scids: Distributed Soft Computing Intrusion Detection System" Journal of Network and Computer Applications, Pp 1-19.
Anish Das and S. Siva Sathya, 2012. "A Fuzzy Based Divide And Conquer Algorithm For Feature Selection In Kdd Intrusion Detection Dataset" International Journal Of Computer Science And Informatics, Vol-2, Pp 46-50.
Bapuji V., R. Naveen Kuma and Dr. A. Govardhan, S.S.V.N. Sarma, 2012. "Soft Computing And Artificial Intelligence Techniques For Intrusion Detection System" Network And Complex Systems, Vol-2, Pp 24-33.
Chadrashekhar A.M. and K. Raghuveer, 2013. "Intrusion detection technique by using K-means, Fuzzy Neural network and SVM classifier" (ICCCI), Pp 450-456.
Devaraju S. and Dr. S.Ramakrishnan 2011. "Performance Analysis Of Intrusion Detection System is Using Various Neural Network Classifiers" In IEEE International Conference On Recent Trends In Information Technology, Icrtit, Pp 1033-1038.
Dewan Md. Farid, Jerome Darmont, Nouria Harbi, Nguyen Huu Hoa and Mohammad Zahidur Rahman, 2008. "Adaptive Network Intrusion Detection Learning: Attribute Selection And Classification" Pp 1-5.
Gary Stein, Bing Chen, Annie S. Wu and Kien A. Hua, 2011. "Decision Tree Classifier For Network Intrusion Detection With Ga-Based Feature Selection" Pp 1-6.
Hachmi Fatma and Limam Mohamed, 2013. "A Two-Stage Technique To Improve Intrusion Detection Systems Based On Data Mining Algorithms" IEEE, Pp 1-6.
Jonatan Gomez and Dipankar Dasgupta, 2002. "Evolving Fuzzy Classifiers for Intrusion Detection" In IEEE.
Jongsuebsuk P., N. Wattanapongsakorn and C. Charnsripinyo, 2013. "Network Intrusion Detection With Fuzzy Genetic Algorithm For Unknown Attacks" IEEE, (Icoin) Pp 1-5.
Li Liu, Pengyuan Wan, Yingmei Wang and Songtao Liu 2014. "Clustering And Hybrid Genetic Algorithm Based Intrusion Detection Strategy "Telkomnika Indonesian Journal of Electrical Engineering, Vol.12, Pp 762-770.
Mohammad A. Faysel and Syed S. Haque, 2010. "Towards Cyber Defense: Research In Intrusion Detection And Intrusion Prevention Systems" Ijcsns International Journal Of Computer Science And Network Security, Vol.10, Pp 316-325.
Morteza Amini, Rasool Jalili and Hamid Reza Shahriari, 2006. "Rt-Unnid: A Practical Solution To Real-Time NetworkBased Intrusion Detection Using Unsupervised Neural Networks" Computers \& Security, Pp 459-468.
Ren Hui Gong, Mohammad Zulkernine and Purang Abolmaesumi, 2005. "A Software Implementation of A Genetic Algorithm Based Approach To Network Intrusion Detection" IEEE, Pp 250-258.
Shafigh Parsazad, Ehsan Saboori and Amin Allahyar, 2012. "Fast Feature Reduction In Intrusion Detection Datasets" Mipro, Pp 1023-1029.
Srilatha Chebrolu, Ajith Abraham and Johnson P. Thomas, 2005. "Feature Deduction And Ensemble Design of Intrusion Detection Systems" Computers \& Security, Pp 295-307.
Tansel Ozyer, Reda Alhajj and Ken Barker, 2007. "Intrusion Detection By Integrating Boosting Genetic Fuzzy Classifier And Data Mining Criteria For Rule Pre-Screening" Journal Of Network and Computer Applications, Pp 99-113.

Venkatachalam V. and S. Selvan 2007. "Intrusion Detection Using An Improved Competitive Learning Lamstar Neural Network" In Ijcsns International Journal Of Computer Science And Network Security, Vol.7, Pp 255-264.

Zorana Bankovic, Dusan Stepanovic, Slobodan Bojanic and Octavio Nieto-Taladriz, 2007. "Improving Network Security Using Genetic Algorithm Approach" Computers and Electrical Engineering, Pp 1-14.


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