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# Bayesian Classifier Approach For An Effective Outlier Intrusion Detection System In Mobile Adhoc Networks

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Abstract - Networks are protected using many firewalls and encryption software's. But many of them are not sufficient and effective. Most intrusion detection systems for mobile ad hocnetworks are focusing on either routing protocols or itsefficiency, but it fails to address the security issues. Some of the nodes may be selfish, for example, by not forwarding the packets to the destination, thereby saving the battery power. Someothers may act malicious by launching security attacks likedenial of service or hack the information. The ultimate goal of the security solutions for wireless networks is to provide security services, such as authentication, confidentiality, integrity, anonymity, and availability, to mobile users. This paper incorporates agents and data mining techniques to prevent anomaly intrusion in mobile adhoc networks. Homeagents present in each system collects the data from its ownsystem and using data mining techniques to observed the local anomalies. The Mobile agents monitoring the neighboring nodes and collect the information from neighboring home agents to determine the correlation among the observed anomalous patterns before it will send the data. This system was able to stopall of the successful attacks in an adhoc networks and reduce the false alarm positives.

# Keywords— Mobile agents, Intrusion detectionsystem, Adhoc networks, Network Security.

# 1. INTRODUCTION

Intrusion detection is an important part of computersecurity. It provides an additional layer of defense againstcomputer is use after physical, authentication and accesscontrol [5]. A mobile ad hoc network is a collection ofwireless mobile hosts forming a dynamic networkinfrastructure without any standard infrastructure orcentralized administration. The flexibility in space and timeinduces new challenges towards the security infrastructure. The

nature of mobility creates new vulnerabilities due to theopen medium, dynamically changing network topology, cooperative algorithms, lack of centralized monitoring andmanagement points and yet many of turn out to the proven securitymeasures ineffective. Therefore. the traditionalway protecting wired/wireless networks with firewalls and encryption software is no longer sufficient [7]. Military, University campuses and conference settings also gain onaccount of these wireless networks since they easycollaboration efficient allow and

communication on the fly withoutthe need for costly network infrastructure.

Expectations are also high with respect to the use of thesenetworks in places like hotels, airports etc. But a vital problemthat must be solved in order to realize these applications of adhoc networks is that concerning the security aspects of suchnetworks [8]. Intrusion detection is used in the networks bycomparing the set of baselines of the system with the presentbehavior of the system [3]. Thus, a basic assumption is that thenormal and abnormal behaviors of the system can becharacterized.

The intrusion detection community has beendeals mainly on wired networks, but it is lack of security inwireless networks. Anomaly detection and misuse detection or signature detection are the two techniques for intrusiondetection system. used detection describes the abnormal patterns of behavior, where "abnormal" patterns are definedbeforehand. detection relies Misuse on the use of specificallyknown patterns of unauthorized behavior. Thus thesetechniques rely on sniffing packets and using the sniffedpackets for analysis. In order to realize these ID techniques thepackets can be sniffed on each of the end hosts. This is called as host intrusion detection (HID). It is also possible to sniffthese packets on certain predetermined machines in thenetwork. This is called as network intrusion detection (NID).

Mobile agents are a special type of agents defined as processes capable of roaming through large networks such as the adhoc wireless network, interacting with machines, collecting information and returning after executing the tasksadjusted by the user. The nature of mobility in a wirelessnetworks creates an vulnerability due to the open medium, dynamically changing networks. In order to avoid such circumstance, to develop new architecture

and mechanisms toprotect the wireless networks and mobile computing applications [7].

The remainder of the paper is organized as follows, Section2 contains related work, Section 3 describes our proposed approach, and Section 4 describes the Results and conclusion.

# 2. RELATED WORK

Traditional security mechanism such as intrusion detection system, firewall and encryption methods are not sufficient to provide security in an adhoc networks. Countering threats toan organization's wireless adhoc network is an important area of research. Intrusion detection means identifying any set ofactions that attempt to compromise the integrity, confidentiality or availability of resource [3]. Manytechniques have been discussed to prevent attacks in anwireless adhoc networks as follows.

Ricardo Puttini et al [16], propose design and development of the IDS are considered in 3 main stages. A parametricalmixture model is used for from behavior modeling referencedata. The associated Bayesian classification thedetection algorithm [15]. MIB variables are used to provideIDS needed information. Experiments of DoS and scannerattacks validating the model are presented as well. Joã B. D.oCabrera Et al [17], provides the solution of intrusion detectionin Mobile Networks Ad-Hoc (MANETs). utilizing ensemblemethods. A three-level hierarchical system for data collection, processing and transmission is described. Local IDS(intrusion detection systems) are attached to each node of theMANET, collecting raw data of network operation, and computing a local anomaly index measuring the mismatchbetween the current node operation and a baseline normaloperation. The complete suite of algorithms was implementedand tested, under two types of MANET routing protocols andtwo types of attacks against the routing infrastructure.

Yongguang Zhang et al [18], propose intrusiondetection and response mechanisms are developing forwireless ad- hoc networks. The wireless ad-hoc network isparticularly vulnerable due to its features of open medium, dynamic changing topology, cooperative algorithms, lack ofcentralized monitoring and management point, and lack of aclear line of defense. Many of the intrusion detectiontechniques developed on a fixed wired network are notapplicable in this new environment. Farroq et al [19] proposethe signature detection technique and investigate the ability of various routing protocols to facilitate intrusion detectionwhen the attack signatures are completely known. We showthat reactive ad-hoc routing protocols suffer from a seriousproblem due to which it might be difficult to detect intrusionseven in the absence of mobility. Mobility makes the problemof detecting intruders harder.

Vijay Bhuse et al [10], propose lightweight methods todetect anomaly intrusions in wireless sensor networks(WSNs). The main idea is to reuse the already available systeminformation that is generated at various layers of a networkstack. This is the different approach for anomaly intrusiondetection in WSNs.

Hongmei Deng et al [21], propose theunderlying distributed and cooperative nature of wireless adhoc networks and adds one more dimension of cooperation to the intrusion detection process. That is, the anomaly detection is performed in a cooperative way involving the participation of multiple mobile nodes. Unlike traditional signature-basedmisuse detection approaches, the proposed scheme detects various types of intrusions/attacks based on the model learnedonly from normal network behaviors. Without therequirements of pre-labeled attack data, the approacheliminates the time-consuming labeling process and theimpacts of imbalanced dataset.

Bo Sun et al [22], propose we first introduce two differentapproaches, a Markov chain-based approach and a Hotelling'sT2 test based approach, to construct local IDSs for MANETs. Then demonstrate that nodes' moving speed, a commonly usedparameter in tuning IDS performances, is not an effectivemetric to under IDS performances tune different mobilitymodels. To solve this problem, author further propose anadaptive scheme, in which suitable normal profiles and orresponding proper thresholds can be selected adaptively byeach local IDS through periodically measuring its local linkchange rate, a proposed unified performance metric.

Haiguang Chen et al [23], propose lightweight anomalyintrusions detection. In the scheme, author investigates different key features for WSNs and defines some rules tobuilding an efficient, accurate and effective IntrusionDetection Systems (IDSs). We also propose a moving windowfunction method to gather the current activity data. Thescheme fits the demands and restrictions of WSNs. Thescheme does cooperation among need any monitor nodes. Simulation results show that the proposed IDSs are efficientand accurate in detecting different kinds of attacks.

Gabriela F. Cretu et al [24], propose the use of modelexchange as a device moves between different networks as ameans to minimize computation and traffic utilization. Anynode should be able to obtain peers' model(s) and evaluate itagainst its own model of "normal" behavior.

Yu Liu et al [25], propose game theoretic framework to analyze the interactions between pairs of attacking/defendingnodes using a Bayesian classifier formulation.

We study the achievableNash equilibrium for the attacker/defender game in both staticand dynamic scenarios. The dynamic Bayesian game is a more realistic model, since it allows the defender to

consistentlyupdate his belief on his opponent's maliciousness as the gameevolves. A new Bayesian hybrid detection approach issuggested for the defender, in which a lightweight monitoring system is used to estimate his opponent's actions, and aheavyweight monitoring system acts as a last resort of defense.

Many authors proposed different techniques to preventattacks in wireless adhoc networks. But all these methodsreported to have a lot of pros and cons of its own proposal. Theauthors mainly classified their mechanism as signature method, anomaly method. In Signature Based Method, a threat isalways be stored in database. New threat being discovered in the wild and the signature for detecting that threat. ThisMechanism would be unable to detect the new threat. In Anomaly Based Method, it monitors system and its networkbehaviors. It set the baseline of network and system. Thismechanism work effectively against wireless networks but itgenerates some false positive results. In this paper, a newattempt has been made and worked out effectively againstattacks in wireless networks. This paper incorporates agentand data mining method to provide solution against securityissues in MANET networks. With the help of home agent andmobile agents, it gathers information from its own system andneighboring system to any attack and through datamining techniques to find out the attacks has been made in thatnetworks.

#### 3. PROPOSED APPROACH

Our approach is entirely based on anomaly based method, which has been used to address security problems related toattacks in a wireless networks. This paper incorporates newmethodology such as mining and agents to provide solutions against wireless networks. Our Proposal provides the three different techniques to provide suffice security solution tocurrent node, Neighboring Node and

Global networks. hefollowing figure clearly depicts the architecture of the systemto prevent the attacks in wireless networks. he following section outlines each module's work in detail.

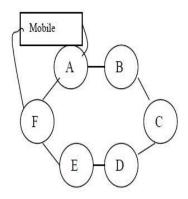


Fig 1: Proposed System Architecture Outlines

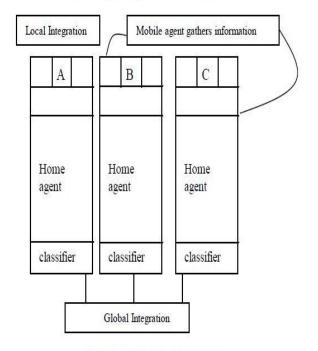


Fig 2: Proposed System Architecture

# A. Home agent

Home agent is present in each system and it gathersinformation about its system from application layer to routinglayer. Our proposed system provides solution in threetechniques. 1. It monitors its own system and its environmentdynamically. It uses classifier construction to find out the local anomaly. 2.

Whenever the node want to transfer theinformation from the node F to B. It broadcast the message to E and A. Before it sends the message, it gathers theneighboring nodes (E &B) information using mobile agent. It calls the classifier rule to find out the attacks with help of testtrain data. 3. It provides same type of solution throughout the global networks. It has been explained in the following section.

- 1) Current node Home Agent is present in the system and itmonitors its own system continuously. If an attackersends any packet to gather information or broadcastthrough this system, it calls the classifier construction to find out the attacks. If an attack has been made, it willfilter the respective system from the global networks.
- 2) Neighboring node Any system in the network transferany information to some other system, it broadcastthrough intermediate system. Before it transfer themessage, it send mobile agent to the neighboring node and gather all the information and it return back to the system and it calls classifier rule to find out the attacks. If there is no suspicious activity, then it will forward themessage to neighboring node.
- 3) Data collection Data collection module is included foreach anomaly detection subsystem to collect the values of features for corresponding layer in an system. Normal profile is created using the data collected duringthe normal scenario. Attack data is collected during the attack scenario.
- **4) Data preprocess** The audit data is collected in a file and it is smoothed so that it can be used for anomaly detection. Data preprocess is a technique to process the information with the test train data. In the entire layer anomaly detection systems, the above mentioned preprocessing technique is used.

# B. Cross feature analysis for classifier sub model

#### construction

- 1) Each feature or character vector f in the training data set, calculate classifier C, for each feature fi using  $\{f_1, f_2... f_i\text{-I}, f_i\text{+.-}f k\}$   $C_i$  is learned from the training data set using Navy Bayesian classification algorithm. The probability  $P(f_{il}f_l, f_2 ..., I f_{i^-} i, f+1, ..., f_k)$  is learned.
- 2) Compute the average probability for each feature vector f, and save in a probability distribution matrix M. A decision threshold 0 is learned from the training data set. Normal profile is created using the threshold value. If the probability is greater than threshold value it is labeled as normal, otherwise it is labeled as abnormal. Anomaly detection.

**Input:** Preprocessed train data, preprocessed test data **Output:** Percentage of anomaly

- 1) Read processed data set file
- 2) Call Bayesian classifier program for training the classifier for anomaly detection
- 3) Read the test data file
- 4) Test the classifier model with the test data file
- 5) Print the confusion matrix to show the actual class vs

Predicted class

6) Percentage of anomaly is calculated as followsPercentage = Number of predicted abnormal classX

100Total number oftraces.

# C. Local integration

Local integration module concentrate on self system and itfind out the local anomaly attacks. Each and every system under hat wireless networks follows the same methodology toprovide secure global networks.

# **D.** Global integration

Global integration module is used to find the intrusionresult for entire network. The aim of global integration is toconsider the neighbor node(s) result for taking decisiontowards response module.

# 4. RESUTLTS AND DISCUSSION

There are many number of attacks has been tested to preventattacks in wireless network. This system not only blocks theapplication oriented issues and it stops some of the networksecurity issues. Consider this limited number of attacks andtested with this proposed system to find out the attacks and gota encourage results. These are the parameters has been take toanalyze the proposed system to find out the efficiency.

TABLE 1: INPUT PARAMETER CONSIDERATION

Parameters	Values
No of nodes	30
Terrain range	2000x 2000 Meters
Routing layer protocol	DSR
Mobility model	Random way point

This system can tested with limited number of attackspresent in the wireless networks. It shows the encouragementresults to support the proposed system. Detection rate of anomaly rate in our proposed system is high and it encourages the system.

TABLE 2: DETECTION RATE OF THE PROPOSED SYSTEM

Detection Module	Detection Rate
Anomaly Detection (A)	80%
Local Integration (D)	95.41%
Global Integration (E)	94.33%

This system act as an Intrusion prevention system to detectand prevent the attacks. But the drawback of existing Intrusionprevention system can generate the more false alarms, but itmay work efficiently. This system can able to stop the attacksas well as it could not generate the false alarms and it workeffectively against the web parameter attacks. Consider thislimited number of access and tested with this proposed system to find out the alarm rates.

TABLE 3: ALARM RATE OF THE PROPOSED SYSTEM

Detection Module	False positivity
Anomaly Detection	1.0%
Local Integration (D)	0.8%
Global Integration	0.75%

# 5. CONCLUSION

In this work, an anomaly detection system comprises ofdetection modules for detecting anomalies in each layer. This system is cooperative and distributive; it considers theanomaly detection result from the neighbor node(s) and sendsthe current working node's result to its neighbor node(s). Experimental results show that detection rate is increasedwhen compared to the other mechanism. False positive rate isalso mechanism. reduced in this Traditional securitymechanism such as IDS and firewall have not been sufficientto provide the security of wireless networks, however, thismechanism is able to block abnormal approach to wirelessnetworks and to detect previously unknown attacks as well asvariations of

known attacks.

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