

GROUP MODELLING OF SECURITY ALERT AND ANALYSIS FOR INTRUDER DETECTION IN ROBOTIC NETWORK

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Abstract:

Multi-agent systems are currently applied to solve complex problems. The security of networks is an eloquent example of a complex and difficult problem. A new model-concept Hybrid Sensitive Robot Metaheuristic for Intrusion Detection is introduced in the current paper. The proposed technique could be used with machine learning based intrusion detection techniques. The new model uses the reaction of virtual sensitive robots to different stigmergic variables in order to keep the tracks of the intruders when securing a sensor network.

Keywords: intrusion detection, sensor network, intelligent agents.

1. INTRODUCTION

Prevention and detection of intruders in a secure network is nowadays a challenging issue. The intrusion detection system based on computational intelligence (CI) has proved in time to have huge advantages over traditional detection systems due to characteristics of CI methods: adaptation, fault tolerance, high computational speed etc. It is essential to design efficient Intrusion Detection Systems (IDS) especially for open medium networks as wireless sensor devices. The intrusions could be misclassified intrusions and anomaly intrusions. Misclassified intrusions are the attacks knowing the weak points of a system. Anomaly intrusions are based on observations of normal system usage patterns and detecting deviations from the given norm. The mentioned intrusions are hard to quantify because there are no fixed patterns that can be monitored and as a result a more fuzzy approach is often required. The Intrusion Preventing Systems (IPS) are network security appliances that monitor network and/or system activities for malicious activities. IPS is a device used to block all the unwanted access to the targeted host, to remove malicious part of packets and as well it may re-configure the network device where an attack is detected [3]. The aim of the current paper is to provide an effective stigmergic-based technique for IDS in a sensor network graph, that consist of multiple detection stations called sensor nodes. The new Hybrid Sensitive Robot Metaheuristic for Intrusion Detection (HSRM-ID) model uses a collection of robots endowed with a stigmergic sensitivity level. The sensitivity of robots allow them to detect and react to different stigmergic variables involving the attacks into a secure network. The hybrid model combines elements from Sensitive Robot Metaheuristic (SRM) [19] as Ant Colony System (ACS) [10], autonomous mobile robots and the intrusion detection based on emotional ants for sensors (IDEAS).

2. STIGMERGIC ROBOTS

The metaheuristic Sensitive Robot Metaheuristic combining the concepts of stigmergic communication and autonomous robot search is used to solve NP-hard optimization problems. An

important characteristic of stigmergy is that individual behavior modifies the environment, which in turn modifies the behavior of other individuals [11]. The SRM technique attempts to address the coupling between perception and action as direct as possible in an intelligent stigmergic manner. As it is known, robot communication relies on local environmental modifications that can trigger specific actions. The set of the rules defining actions (stimuli pairs) used by a homogeneous group of stigmergic robots defines their behavior and determines the type of structure the robots will create. Robot stigmergic communication does not rely on chemical deposition as it is for artificial ant-based colonies [10]. A stigmergic robot action is determined by the environmental modifications caused by prior actions of other robots. The value of quantitative stigmergy modify the future actions of robots. Discrete stimulus are involved in qualitative stigmergy and the action is switched to a different action [4, 26]. Some real-life applications of the behavior-based approach, including autonomous robots, are in data mining, military applications, industry and agriculture, waste management, health care. Lee et al. [15] introduced a data mining classification mechanism with association rules from the audit data - knowledge present in a knowledge base - providing gaudiness for data gathering and feature selection. In order to detect abnormal behavior one can use genetic algorithms, see for example [1]. In [18], neural networks use back propagation MLP for a small network in order to detect anomalies and identify user profiles after end of each log session. The proposed (HSRM) can be modelled using two distinct groups of sensitive stigmergic robots. The first group of robots-agents is endowed with small sensitive values SSL and they are sensitive-explorers (sSSL: small SSL-robots). They can sustain diversification in intruders searching. In the second group are the robots-agents with high sensitive stigmergic values (hSSL: high SSL-robots).

3. NEW CONCEPT

In the following is performed an analyze of the Hybrid Sensitive Robot Algorithm for Intrusion Detection. The artificial pheromone from the edges of the sensor network graph reveals as the attacked zone within the network. Each bio-inspired robot uses his one specific properties as his level of sensitivity in order to detect the intruders and the artificial stigmergy in order to find the attacked edges. Table 1 illustrates the behavior of different groups of sensitive bio-inspired virtual robots when investigate the sensor network in search of intrusion. As a concept, the introduced model Hybrid Sensitive Robot Algorithm for Intrusion Detection has more chances to improve the intrusion detection systems comparing with the existing approaches from the literature, due to the sensitivity property of the bio-inspired robots. As well the diversity of robots groups implies also different values of virtual pheromone trail values. The robots with small stigmergic value are constantly sustaining diversification in intruders searching and as a complementary action, the robots with high sensitive stigmergic values are testing the already identified networks attacked regions. In the future we will perform numerical experiments to assess the performance of the proposed algorithm.

4. INTRUSION DETECTION TECHNIQUES USING ARTIFICIAL INTELLIGENCE

At first are introduced the main concepts of IDS followed by a survey of Artificial Intelligence-based existing models for computer security. Intrusion detection technology is a technology designed to monitor computer activities for the purpose of finding security violations. IPS is able to detect and prevent attacks but it has not deeper detection capabilities of IDS. Neither of Intrusion Detecting System and Intrusion Prevention System is capable to provide in depth security. Intrusion Detecting and Prevention System I(IDPS), a combinations of IDS and IPS, is a more effective system capable of detection and prevention. An example of network-based system is Snort [14]. Snort is an open source network intrusion prevention and detection system - nowadays a standard for IPS - that combines the

benefits of signature, protocol and anomaly-based inspection. host-based systems describe the class of software able to monitor a single system, analyse characteristics and log to at one host. These systems are deployed on critical hosts. wireless-based systems analyse wireless traffic to monitor intrusion or any suspicious activity. They scan traffic but are not able to identify attack in the application layer or higher layer network protocols as UDP and TCP. It may be deployed at the point where unauthorized wireless network could be accessed. According to Beg et al. [3], the intrusion detection classical algorithms have the following disadvantages: false alarm rate and constant updates of database with new signatures. The network administrator responds to alarms and updates the signatures that increases in time. For example, in the already mentioned Snort signatures increased from 1500 to 2800 over two years [14]. In order to improve the administrator work, reducing the number of false alarms and better intrusion detection are introduced artificial intelligence mechanisms [23]. Some of AI techniques used in intrusion detection are data mining, genetic algorithm, neural network, multi-agents, ant-net miner, etc. Lee et al. [15] introduced a data mining classification mechanism with association rules from the audit data - knowledge present in a knowledge base - providing guidance for data gathering and feature selection. In order to detect abnormal behavior one can use genetic algorithms, see for example [1]. In [18], neural networks use back propagation MLP for a small network in order to detect anomalies and identify user profiles after end of each log session. It shall also be remarked that several of the leading methods for detecting intrusions and detecting intrusions are hybrid artificial approaches, which combine different AI solution techniques [9, 16, 25]. Some hybrid methods used in the literature are data mining and fuzzy logic techniques [16], data mining and genetic algorithm selecting the best rules for the system [9]. In the future could be implemented hybrid models involving intelligent evolutionary agents [12] and dynamic decision boundary using Support Vector Machine [24] for handle a large number of features. Banerjee et al. [2] introduced an intrusion detection based on emotional ants for sensors (IDEAS), which could keep track of the intruder trails. This technique is able to work in conjunction with the conventional machine learning based intrusion detection techniques to secure the sensor networks.

5. HYBRID SENSITIVE ROBOT METAHEURISTIC FOR INTRUSION DETECTION

In this section we introduce a new hybrid metaheuristic in order to detect the intruders in a sensor network. The new model is called Hybrid Sensitive Robot Metaheuristic for Intrusion Detection (HSRM-ID), is based on Sensitive Robot Metaheuristic (SRM) introduced in [19] and uses a specific rule in order to generate a state of thinking or the choice of an intruder [2]. The proposed (HSRM) can be modelled using two distinct groups of sensitive stigmergic robots. The first group of robots-agents is endowed with small sensitive values SSL and they are sensitive-explorers (sSSL: small SSL-robots). They can sustain diversification in intruders searching. In the second group are the robots-agents with high sensitive stigmergic values (hSSL: high SSL-robots). They are sensitive-exploiters and could exploit intensively the regions already identified with attacks from intruders. In time, based on the experience of robots-agents, the sensitive stigmergic level SSL can increase or decrease. The pseudo-code description of the Hybrid Sensitive Robot Metaheuristic for Intrusion Detection is described in what it follows. The stigmergic value of an edge is τ and the visibility value is η . A tabu list with the already visited nodes is maintained, see [10] for more details. In order to divide the colony of m robots in two groups it is used a random variable uniformly distributed over $[0,1]$. Let q be a realization of this random variable and q_0 a constant $0 \leq q_0 \leq 1$. If the inequality $q > q_0$ stands the robots are endowed with small sensitive stigmergic value sSSL robots and otherwise they are highly sensitive stigmergic robots (hSSL). A hSSL-robot uses the information supplied by the sSSL robots.

CONCLUSION

Nowadays the networks are threatened by security attacks and resource limitations. In order to deal with this security network problem efficient intruders detection and prevention systems are used. Within this paper we introduce a new concept Hybrid Sensitive Robot Algorithm for Intrusion Detection based on bio-inspired robots. It is used a qualitative stigmergic mechanism, each robot is endowed with a stigmergic sensitivity level facilitating the exploration and exploitation of the search space. In the future some computational tests will be proposed and further hybrid AI techniques will be involved for securing the networks.

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