HYBRID INTELLIGENT APPROACH FOR NETWORK INTRUSION DETECTION

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ABSTRAK

Sejak kebelakangan ini, rangkaian komputer telah meluas dan sangat rumit. Banyak informasi sensitif disalurkan kepada pelbagai jenis peranti komputer, dari komputer mini ke pelayan dan juga dari komputer mini ke peranti mudah alih. Perubahan ini menyebabkan serangan ke atas maklumat penting ke atas sistem rangkaian semakin bertambah setiap tahun. Pencerobohan adalah ancaman utama terhadap rangkaian. Ia ditakrifkan sebagai satu siri aktiviti yang bertujuan untuk menjejaskan keselamatan sistem rangkaian dari segi kerahsiaan, integriti dan ketersediaan. Oleh itu, pengesanan pencerobohan adalah sangat penting sebagai sebahagian daripada pertahanan. Oleh itu, ia perlu meningkatkan teknik pengesanan pencerobohan rangkaian dan sistem. Disebabkan pendekatan pengesanan pencerobohan sebelum ini terlalu terhad, kami mencadangkan pendekatan hibrid pintar untuk mengesan pencerobohan rangkaian berdasarkan pengelompokan k-means algoritma dan mesin sokongan vektor algoritma. Tujuan penyelidikan adalah untuk mengurangkan kadar penggera palsu dan juga untuk meningkatkan kadar pengesanan untuk dibandingkan dengan pendekatan pengesanan pencerobohan yang sedia ada. Pencerobohan dataset NSL-KDD telah digunakan untuk latihan dan menguji pendekatan yang dicadangkan. Beberapa langkah telah dilakukan sebelum tujuan pengelasan untuk meningkatkan prestasi pengelasan. Pertama, menyatukan jenis dan menapis dataset melalui data transformasi. Kemudian, pemilihan ciri algoritma telah digunakan untuk membuang ciri yang tidak relevan dalam tujuan pencerobohan. Ciri-ciri yang terpilih telah mengurangkan 41 ciri kepada 21 ciri untuk mengesan pencerobohan dan kemudian kaedah-kaedah yang biasa digunakan untuk dilaksanakan serta mengurangkan perbezaan di antara maklumat. Pengelompokan adalah langkah terakhir pemprosesan sebelum pengelasan dijalankan menggunakan kmeans algoritma. Klasifikasi telah dilakukan dengan menggunakan mesin sokongan vektor. Selepas latihan dan menguji pendekatan pintar hibrid yang telah dicadangkan, keputusan penilaian prestasi telah menunjukkan bahawa ia mencapai ketepatan yang tinggi dan kadar pengesanan palsu yang rendah. Ketepatannya ialah 96.025% dan penggera palsu adalah 3.715%.

Kata Kunci: Rangkaian Pengesanan Pencerobohan, Pendekatan Pintar Hibrid, Rangkaian Serangan, Pengelompokan, Klasifikasi, Pencerobohan dataset NSL-KDD, K-Means algoritma, Mesin Sokongan Vektor algoritma.

ABSTRACT

In recent years, computer networks are broadly used, and they have become very complicated. A lot of sensitive information passes through various kinds of computer devices, ranging from minicomputers to servers and mobile devices. These occurring changes have led to draw the conclusion that the number of attacks on important information over the network systems is increasing with every year. Intrusion is the main threat to the network. It is defined as a series of activities aimed for exposing the security of network systems in terms of confidentiality, integrity and availability, as a result; intrusion detection is extremely important as a part of the defense. Hence, there must be substantial improvement in network intrusion detection techniques and systems. Due to the prevailing limitations of finding novel attacks, high false detection, and accuracy in previous intrusion detection approaches, this study has proposed a hybrid intelligent approach for network intrusion detection based on k-means clustering algorithm and support vector machine classification algorithm. The aim of this study is to reduce the rate of false alarm and also to improve the detection rate, comparing with the existing intrusion detection approaches. In the present study, NSL-KDD intrusion dataset has been used for training and testing the proposed approach. In order to improve classification performance, some steps have been taken beforehand. The first one is about unifying the types and filtering the dataset by data transformation. Then, a features selection algorithm is applied to remove irrelevant and noisy features for the purpose of intrusion. Feature selection has decreased the features from 41 to 21 features for intrusion detection and later normalization method is employed to perform and reduce the differences among the data. Clustering is the last step of processing before classification has been performed, using k-means algorithm. Under the purpose of classification, support vector machine have been used. After training and testing the proposed hybrid intelligent approach, the results of performance evaluation have shown that the proposed network intrusion detection has achieved high accuracy and low false detection rate. The accuracy is 96.025 percent and the false alarm is 3.715 percent.

Keywords: Network Intrusion Detection, Hybrid Intelligent Approach, Network Attacks, Clustering, Classification, NSL-KDD intrusion dataset, K-Means algorithm, Support Vector Machine algorithm.

DEDICATION

Every challenging work needs self-efforts as well as guidance and support of others, especially those who are very close to our heart.

Therefore, I dedicate this humble work

To my sweet and beloved

FAMILY

Whose love, support, and pray of day and night for making me able to reach such success.

To my lovely homeland

IRAQ

Which opening my eyes to this world. I hope it will get the peace soon.

To the marvelous land

MALAYSIA

Which granted me the opportunity to complete my study.

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Monday, October 20, 2014

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LIST OF ABBREVIATIONS

A Accuracy

AI Artificial Intelligent

ANN Artificial Neural Network

DoS Daniel of Services Attack

DR Detection Rate

FAR False Alarm Rate

FN False Negative

FP False Positive

HIDS Host-based Intrusion Detection System

IDS Intrusion Detection System

NID Network Intrusion Detection

NIDS Network-based Intrusion Detection System

R2L Remote to Local Attack

SVM Support Vector Machine

TN True Negative

TP True Positive

U2R User to Root Attack

CHAPTER ONE

INTRODUCTION

1.1 Introduction

This chapter has discussed briefly the background of network security impacts, network intrusion problems and its solutions. On the other hand, it presents amply the statement of the problem in this study. This chapter defines the research questions, objectives of this study, the scope of research, research's significance and contributions of the study as well.

1.2 Background of study

In recent years, computer networks are broadly omnipresent and have become very complicated. Almost everybody with a computer or mobile device, is linked with the Internet in order to have access to data or send messages. A lot of sensitive information passes through various kinds of computer devices, ranging from minicomputers to servers and mobile devices (Elbasiony et al., 2013; Upadhyaya & Jain, 2013). In a wide scale, all governments, higher education organizations and different organizations depend on the network computer systems for the daily processes to perform, and network computer system play an essential role for the processes (Shanmugam & Idris, 2011).

The contents of the thesis is for internal user only

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