

# INTRUSION DETECTION SYSTEMS USING K-MEANS CLUSTERING SYSTEM

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Bachelor of Computer Science (Software  
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## **ABSTRACT**

Internet is the biggest platform for people all over the world to connect with each other, and to search for important information and such. Along with the raising of internet usage, the number of cases of intrusion attacks also increases. Because of this, intrusion detection is important, especially for large companies which held huge and confidential data and information. This system works to detect the abnormal connection of a network so that a stronger protection could be build. Since the attack is not restricted into only one type, a data mining technique is applied to classify all types of attack from a huge amount of data entering into a network. By using this technique, the intrusion detection system could work better. In this research, data mining technique of K-means clustering system are used to detect the intrusion and attack. 1999 KDD Cup Dataset is used for training, testing, and validation of the system. The dataset is famous among intrusion detection system researcher for its data which resembles real attacks at real times.

## ABSTRAK

Internet ialah platform terbesar yang digunakan oleh segenap dunia untuk berhubung dengan satu sama lain, dan mencari maklumat penting dan terkini. Dengan peningkatan jumlah penggunaan internet, jumlah kes-kes yang melibatkan serangan pencerobohan juga turut meningkat. Disebabkan oleh hal ini, pengesanan pencerobohan adalah penting, terutama sekali untuk syarikat-syarikat besar yang mempunyai data dan maklumat yang banyak dan sulit. Sistem ini berfungsi untuk mengesan sambungan rangkaian yang abnormal supaya satu sistem perlindungan yang lebih kuat boleh dibina. Oleh kerana serangan pencerobohan tidak terhad kepada satu jenis sahaja, teknik perlombongan data diaplikasikan untuk mengklasifikasikan semua jenis serangan daripada jumlah data yang banyak yang memasuki sesebuah rangkaian. Dengan menggunakan teknik ini, sistem pengesanan pencerobohan boleh berfungsi dengan lebih baik. Dalam kajian ini, teknik perlombongan data yang digunakan ialah sistem pengelompokan K-means digunakan untuk mengesan pencerobohan dan serangan. Set data yang digunakan untuk latihan, ujian, and pengesanan sistem adalah 1999 KDD Cup Dataset. Set data ini adalah terkenal dikalangan pengkaji pengesanan pencerobohan kerana data nya adalah menyerupai serangan yang sebenar di masa yang sebenar.

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## LIST OF ABBREVIATIONS

CFS	Correlation-based feature selector
DARPA	Defence Advanced Research Projects Agency
DOS	Denial of Service
EM	Expectation Maximum
FN	False Negative
FP	False Positive
IDS	Intrusion Detection System
KDD	Knowledge Discovery in Database
KNN	K-Nearest Neighbor
MAP	Maximum a Posteriori
R2L	Remote to Local
TN	True Negative
TP	True Positive
U2R	User to Remote

# CHAPTER 1

## INTRODUCTION

### 1.1 INTRODUCTION

Nowadays, internet is the main means of connection between people around the world. Almost everywhere in the world have internet connection, whether it is as wired or wireless. Without proper use of the internet, one could be exposed to the risk of having cyber intruders attacking them. This would let the perpetrators to gain personal data or company confidential illegally. There are a lot of ways for the intruders to intrude such as unauthorized access, denial of service and sending of viruses. In Malaysia, there was a total of 5,078 cybercrimes incidences reported in between January and August 2018 based on Cybercrime Malaysia statistics. Fraud and intrusion each has the total of 2,907 and 699 cases, thus making the two incidents top the charts of MyCERT's statistic.

Intrusion Detection System (IDS) is a system for the users to detect the anomalies of their internet network traffic. There are two categories of IDS which are anomaly detection and misuse detection. Misuse detection will match the computer activity with the stored signatures of the attacks the systems have known. Anomaly detection is a method to detect intrusions by learning the attribute of normal activity, and it will detect anything that is different from the normal activity.

## **1.2 PROBLEM STATEMENT**

Cybercrimes incidents reached thousands of reports every year. This shows that most of the internet users are still not aware about the problems of the cybercrime and cyber intruder around them. This lack of awareness might be because the users do not know how to check their network traffic activity, and thus, making them vulnerable to the attack of the intruder. Some users would also think that they would never have any intruder because of their position of not being someone important or somebody with high rank. In reality, all internet users should be able to detect their own network traffic activity.

In order to help internet users for this reason, we proposed a research of Intrusion Detection System which is based on misuse and anomaly detection systems. The system would be easy to use, thus making anyone can learn it. This would help them to know whether they are being attacked or not, and be able to find a solution to tighten their cyber security.

## **1.3 OBJECTIVE**

The main objective of this research is to find the right systems to detect the abnormal network traffic activity using IDS. The objectives are:

- i. To design an effective system for cyber intrusion detection to detect abnormal network traffic activity.
- ii. To develop Intrusion Detection System using K-Means clustering algorithm.
- iii. To test the findings and performance of the proposed Intrusion Detection System.

## **1.4 SCOPE**

The scope of this research is divided into several categories, which are;

- i. Dataset
  - a. The dataset that will be used in this research is the 1999 KDD Dataset Intrusion Detection Evaluation Data.
- ii. Language
  - a. The programming language to implement the algorithm used is Java.

## **1.5 THESIS ORGANIZATION**

This thesis consists of five chapters. Chapter 1 will discussed about the introduction of the research, with its problem statement, objectives, and scope. In this chapter, the definition of intrusion detection systems is explained.

Chapter 2 discusses about the literature review. This consists of the review of related works, and to choose the appropriate algorithm to be used in the system.

Chapter 3 is about the methodology to be used in this work. A proposal to solve the existing problem and the flow of conducting the research is being discussed in this chapter.

Chapter 4 shall discuss the implementation on the proposed algorithm towards the dataset. The results of the implementation will be analyzed and discussed.

Chapter 5 is the summary of this research. A summarized version about the experiments and the constraints of the experiment are concluded in this chapter.

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