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NETWORK PROBLEMS DETECTION AND CLASSIFICATION BY ANALYZING SYSLOG DATA



Supervisors:
Dr. Mohammad Shamrie Sainin
Assoc. Prof. Hatim Tahir

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ABSTRAK

Rangkaian penyelesaian masalah adalah satu proses penting yang mempunyai bidang penyelidikan yang luas. Langkah pertama dalam prosedur penyelesaian masalah adalah mengumpul maklumat untuk mengenal pasti permasalahan. Mesej syslog yang dihantar oleh hampir semua peranti rangkaian mengandungi sejumlah besar data yang berkaitan dengan masalah rangkaian. Banyak kajian yang dijalankan sebelum ini didapati telah menggunakan menganalisis data syslog yang boleh membimbing untuk masalah rangkaian dan sebab-sebabnya. Mengesan masalah rangkaian akan menjadi lebih efektif jika masalah yang hendak dikesan telah dikelaskan dari segi lapisan rangkaian. Pengelasan data syslog perlu mengenal pasti mesej syslog yang menghuraikan masalah rangkaian untuk setiap lapisan, dan mengambil kira format yang berbeza dari pelbagai syslog untuk peranti vendor. Kajian ini menyediakan kaedah untuk mengelaskan mesej syslog yang menunjukkan masalah rangkaian dari segi lapisan rangkaian. Alat pengenalpastian data kaedah digunakan untuk pengelasan mesej syslog manakala penerangan bahagian atas mesej syslog telah digunakan untuk proses pengelasan. Apabila mesej syslog berkaitan telah dikenal pasti; ciri kemudiannya dipilih untuk melatih penjodoh bilangan. Enam algoritma pengelasan telah dipelajari iaitu LibSVM, SMO, KNN, Naive Bayes, J48, dan Random Forest. Satu set data sebenar yang diperoleh daripada peranti rangkaian Universiti Utara Malaysia (UUM) digunakan untuk peringkat ramalan. Keputusan merumuskan bahawa SVM menunjukkan prestasi terbaik semasa peringkat latihan dan ramalan. Kajian ini menyumbang pada bidang penyelesaian masalah rangkaian, dan pengelasan.

Keywords data teks: Pengelasan, SVM, Pengesanan Kerosakan..

ABSTRACT

Network troubleshooting is an important process which has a wide research field. The first step in troubleshooting procedures is to collect information in order to diagnose the problems. Syslog messages which are sent by almost all network devices contain a massive amount of data related to the network problems. It is found that in many studies conducted previously, analyzing syslog data which can be a guideline for network problems and their causes was used. Detecting network problems could be more efficient if the detected problems have been classified in terms of network layers. Classifying syslog data needs to identify the syslog messages that describe the network problems for each layer, taking into account the different formats of various syslog for vendors' devices. This study provides a method to classify syslog messages that indicates the network problem in terms of network layers. The method used data mining tool to classify the syslog messages while the description part of the syslog message was used for classification process. Related syslog messages were identified; features were then selected to train the classifiers. Six classification algorithms were learned; LibSVM, SMO, KNN, Naïve Bayes, J48, and Random Forest. A real data set which was obtained from the Universiti Utara Malaysia's (UUM) network devices is used for the prediction stage. Results indicate that SVM shows the best performance during the training and prediction stages. This study contributes to the field of network troubleshooting, and the field of text data classification.

Keywords: Classification, SVM, Fault Detection

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CHAPTER One

INTRODUCTION

Most institutions and organizations, regardless of the business types, rely on networks to manage their business. Any failure or error which occurs in the network will negatively affect their achievements, productivity and services. Therefore, it is necessary to diagnose and detect the reasons behind network failures and problems in order to fix them and reduce similar occurrences in the future. Network troubleshooting, which begins by diagnosing the problems, is a complex process. The first step is to collect information [1].

Collecting information includes answering this question, "what are the potential errors that can lead to network problems and failures?" Kyas [2], has identified five categories of errors: operator error, mass storage problems, computer hardware problems, software problems, and network problems. Hudyma & Fels [3], added two new categories: failure due to denial of service attacks (Worms, Viruses, Trojan Horses and Malicious software), and failure due to disasters such as fire, flood, earthquakes, outages and the like. Network problems, which include hardware and software problems that are directly related to the network architecture [2],account for more than one-third of information technology (IT) failures.

Network architecture is organized as a series of layers or levels [4], Open Systems Interconnection (OSI) model, and Transmission Control Protocol/Internet Protocol (TCP/IP) model, which separates network functionality into modular layers that provide "a common language for network engineers and is usually used in troubleshooting networks."

Troubleshooting could be an efficient process if it relies on a systematic approach which minimizes confusion and shortened troubleshooting time. It is carried out using the Layered Model [5] as problems are normally described in terms of a specific model layer [1]. Network errors could be distributed into the network layers depending on OSI model or TCP\IP model (physical layer, data-link layer, network layer, transport layer, application layer). And knowing the layer that has problems in

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