Indian Institute of Technology Guwahati



System Programming Lab Report

Topic: ADFA-LD Dataset Pre-processing

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1 Introduction

Description of Australian Defence Force Academy-Linux Dataset (ADFA-LD) :

- 1. The dataset was generated on Linux local server running on Ubuntu 11.04, offering a variety of functions such as file sharing, database, remote access and web server.
- 2. Six types of attacks occur in ADFA-LD including two brute force password guessing attempts on the open ports enabled by FTP and SSH respectively, an unauthorised attempt to create a new user with root privileges through encoding a malicious payload into a normal executable, the uploads of Java and Linux executable Meterpreter payloads for the remote compromise of a target host, and the compromise and privilege escalation using C100 webshell. These types are termed as Hydra-FTP, Hydra-SSH, Adduser, Java-Meterpreter, Meterpreter and Webshell respectively.
- 3. 833 and 4373 normal traces are generated for training and validation respectively, over a period during which no attacks occur against the host and legitimate application activities ranging from web browsing to document writing are operated as usual.

2 Task Steps

Following steps have been performed to do the task:

- 1. We split the Attack data for each category into 70% training data and 30% test data. For that, first 7 folders have been used for training data and last 3 for test data. For Normal data, files in "Training_Data_Master" folder have been used for training data and files in "Validation_Data_Master" have been used as test data.
- 2. For each type of attack, the python script first combines all the training data files into one list, then the frequency all the unique n-grams have been found for them as well as for Normal training data files. Concatenating entire training file for a particular class of attack and then running the script on the concatenated file instead of running it individually on each file saves time.
- 3. For finding n-grams, a list of all possible values of n is maintained. For general case, we have included only one element i.e. n=3, the list can be extending just by adding other elements like 5 or 7. And the same program will work because the whole process iterates over all elements of n.

- 4. For calculating the frequencies of all n-grams, the ngrams_freq function first takes three consecutive elements of the list (of concatenated training data files) and makes it a string. In this was, the whole list is iterated and such strings are compared to get the frequency of n-grams. The function returns a dictionary of n-grams frequencies. This also saves our time comparing 3 elements individually. Similarly, the script finds the unique n-grams frequency for Normal files.
- 5. For all the category of attacks and the Normal, top 30% most frequent n-grams are selected. For that, the whole n-gram frequency dictionary is sorted and then top 30% have been picked. These n-grams are used as the features to create the dataset.
- 6. Now, the training dataset is created. File train.csv is created. The frequencies of n-grams which are in features list are found from each and every files which are in training data individually. A row is added in the train.csv file corresponding to each file. Corresponding label has been added for each data.
- 7. Now, for creating the test dataset, first test.csv file is created. And same thing is done on each of the files individually. And one row is added for each file processed.

3 Statistics

For 3-grams:

1. Total number of features: 4142

2. Total size of training dataset: 1338

Distributed as:

AddUser: 56Hydra_FTP: 115Hydra_SSH: 117

Java_Meterpreter: 84Meterpreter: 55

Web_Shell: 78Normal: 833

3. Total size of test dataset: 4613

Number of test data derived from each class:

AddUser: 35Hydra_FTP: 47Hydra_SSH: 59

• Java_Meterpreter : 40

Meterpreter: 20Web_Shell: 40Normal: 4372