Identifying the User As Genuine/Malign Based on Search Logs and Search History

D. Satya Bhavani, P. RajyaLakshmi Sobha Pavani

Abstract—One of the major challenges a developer may face is security issues/threats on the labelled data. The labelled data comprises of system logs, network traffic or any other enriched data with threat/not threat classification. There were few studies which categorized the URLs to a specific category like Arts, Technology, etc. In this paper the main research is on the classification of users based on the search logs(URLs). Manually it is difficult to differentiate the user based on search logs. So, we train a machine learning model that takes raw data as input and classifies the user to genuine or malign. This model helps in intrusion detection/suspicious activity detection. For this first we gather data of past malicious URLs as training set for Naïve Bayes algorithm to detect the malicious users. By implementing KNN algorithm effectively we can detect the malign users up to an accuracy of 94.28%. With the help of Machine Learning algorithms like Naïve Bayes, KNN, Random Forest classifiers we can classify the malign and genuine users.

Keywords: URL, Malign, Naïve Bayes, KNN, Random Forest intrusion detection

I. INTRODUCTION

The main aim of the proposed work is to classify the users based on URLs, who can harm others socially or financially. In this context first we need to understand the web security and phishing. Phishing is a cybercrime in which the attacker try to extract information from the user. To solve that purpose the attacker creates fake and malicious webpages and URLs. When the user frequently visits that page his data is extracted by the attacker and the attacker frequently accesses that page to collect the data. In this proposed work a Machine Learning model is implemented to classify the User based on his search logs or URLs searched. With this trained model we can classify the user to genuine/malign. The classification is done effectively and efficiently to provide the most accurate results. For this we first gather the data of malicious and benign URLs and use it as the training set for machine learning algorithms like Naïve Bayes, KNN to classify the users. The algorithms used are able to classify the users with an accuracy of around 94.28%.

II. METHODOLOGY

A. Architecture

In the proposed work the model takes the input log data and it is pre-processed to find any inconsistent and null data.

B. DATA ACQUISITION

Data acquisition is the process of importing raw data sets into your analytical platform. It can be acquired from traditional databases (SQL and query browsers), remote data (web services), text files (scripting languages), NoSQL storage (web services, programming interfaces), etc. The dataset used in the

Fig.1:System Architecture

Then it is split into training and testing to fit it into the classifier. The Naïve-bayes classifier is trained against training data and is tested using the test data and the accuracy is predicted. Therefore it classifies the user as genuine or malign based on the search logs and search history .The KNN classifier is also implemented for classification of the user to genuine or malign and later the results are compared. The input data is split into train and test using the modules present in the python sklearn module. The Fig.1 represents the system architecture.
The $\mu_y$ and $\sigma_y$ parameters were estimated using maximum likelihood.

**KNN Classifier**

K-Nearest Neighbor is one of the Supervised Machine Learning technique. It assumes the similarity between the new datapoint and the available datapoints and places the new datapoint into the category that is most similar. This algorithm stores all the datapoints and classifies the new data point based on the similarity. K-NN algorithm can be used for both Regression and Classification.

**D. DATA VISUALIZATION**

Data Visualization is the graphical representation of data. This is done using graphs, charts, etc which are used to find the data patterns and outliers present in the data. The project is collected from Kaggle. This data was obtained from different sources of benign and malicious URL’s. The final dataset which serves as the training set for machine learning model consists of around 1780 values.

**C. CLASSIFICATION MODELS**

**Naïve Bayes Classifier**

It is a classification technique which is based on Bayes theorem which assumes the independence of predictors. It assumes that the dependence of one feature in class is not related with the other feature. It can be extended to real-valued attributes by assuming a Gaussian distribution. This extension is called Gaussian Naive Bayes.

**RESULTS**

A. **Testing**

It is a process of executing a program to find the errors if present. If the testing process is done successfully it removes all the errors from the software.

**Table.1.1: Black box Testing**

|  |  |  |
|---|---|---|
| Input | Actual Output | Predicted Output |
| [16,6,324,0,0,0,22,0,0,0,0,0] | 0 | 0 |
| [16,7,263,7,0,2,700,9,10,1153,832,9,2] | 1 | 1 |

B. **Model Accuracy**

The naïve bayes classifier and knn classifier is tested against several test datasets. The models are trained to get a good accuracy. The models trained has the following specifications:

**Fig.4: Naïve Bayes specifications**

```python
1 gnb = GaussianNB()
2 gnb.fit(train_img, train_labels)
```

GaussianNB(priors=None, var_smoothing=1e-09)
KNN Classifier are successfully developed to fit a training data. The trained Gaussian Naïve Bayes model now predicts the user up to an accuracy of 29.14%. The model trained using KNN classifier now predicts the user up to an accuracy of 94.28%. The same is also implemented using Random Forest Classifier which predicts up to an accuracy of 92%. Therefore KNN classifier identifies the user behavior more accurately.

B. Future Scope

A comparison between various machine learning algorithms such as CNN(Convolutional Neural Network), Ada Boost, Bagging, etc can be (TIST), Vol, 4, 2013.

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IV. CONCLUSION AND FUTURE SCOPE

A. Conclusion

performed to get even better results. The model needs to be updated in finite intervals to be more accurate. The more increase the dataset helps in faster suspicious or intrusion detection activity which further increases the security. This can further be extended to obtain the IP address of that user. Identifying the user as genuine/malign based on search logs project is completed successfully. The goal of the project is achieved. Gaussian Naïve Bayes Classifier and