

Mobile Botnet Detection Using Convolutional Neural Networks

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Abstract:

Botnets have been a serious threat to the Internet security. With their constant sophistication and the resilience of them, a new trend has emerged, shifting botnets from the traditional desktop to the mobile environment. As in the desktop domain, detecting mobile botnets is essential to minimize the threat that they impose. Along the diverse set of strategies applied to detect these botnets, the ones that show the best and most generalized results involved is- covering patterns in their anomalous behavior. In the mobile botnet field, one way to detect these patterns is by analyzing the operation parameters of this kind of application. In this paper, we present an anomaly-based and host-based approach to detect mobile botnets. The proposed approach uses machine learning algorithms to identify anomalous behaviors in statistical features extracted from system calls. We were able to test the performance of our approach in a close-to reality scenario. The proposed approach achieved great results, including low false positive rates and high true detection rates. Index Terms—Machine learning, Deep learning, Convolutional Neural Networks (CNN), Botnet Detection, etc.

1. Introduction

Due to the increase of mobile phone users and the increasing number of Android market share nowadays, we noticed that the number of Android malware is also apparently increasing. we extract important features from the Android APK files that can be used to identify a botnet and its family. we find an appropriate machine learning algorithm to classify the Android botnet family with a high recall, and lastly. we developed a system called ABIS (Android Botnet Identification frustum) that includes the identification engine, a web application, and an Android application for the users to check any application before installing it. The rest of this paper is organized as follows.

2. LITERATURE SURVEY

A research paper is a document of a scientific article that contains relevant expertise, including substantive observations, and also references to a specific subject of philosophy and technique.

- 1) ABIS: A Prototype of Android Botnet Identification System:- According to the advanced wire-less technology in nowadays, most people mainly use their mobile phones as an essential tool. At the same time, threats to mobile phones such as viruses, botnets, and other malware are also increasing. However, most users have limited knowledge about mobile threats. Therefore, we would like to reduce the

number of botnet-infected mobile phones before the users install an application on their phones. We developed a system called ABIS (Android

Botnet Identification System) to check Android applications and whether they are possibly malware or not. To identify the Android botnets, our system learns the characteristics of each Android botnet family from the dataset provided by the University of New Brunswick . We analyze the Android APK files, extract important features, and find an appropriate machine- learning technique. As a result, we found that our system can classify the Android botnets with about 96.9 of recall.

- 2) ABC: Android Botnet Classification Using Feature Selection and Classification Algorithms:- Smartphones have become an important part of human lives, and this led to an increase number of smartphone users. However, this also attracts hackers to develop malicious ap- plications especially Android botnets to steal private information and cause financial losses. Due to the fast modifications in the technologies used by malicious application (app) devel- opers, there is an urgent need for more advanced techniques for Android botnet detection. In this paper, a new approach for Android botnet classification based on features selection and classification algorithms is proposed. The proposed approach uses the permissions requested in the Android app as features, to differentiate between the Android botnet apps and benign apps. The Information Gain algorithm is used to select the most significant permissions, and then the classification algorithms Naïve Bayes, Random Forest, and J48 used to classify Android apps as botnet or benign apps. The experimental results show that the Random Forest Algorithm achieved the highest detection accuracy of 94.6 with the lowest false positive rate of 0.099.
- 3) Toward a Detection Framework for Android Botnet:- Android is one of the most popular and widespread operating systems for smartphones. It has several millions of applications that are published at either official or unofficial stores. Botnet applications are a kind of malware that can be published using these stores and downloaded by the victims on their smartphones. In this paper, we propose Android botnet detection method based on a new set of discriminating features extracted from the analysis of Android permissions (i.e. Protection levels for all available Android permissions). Then we compared the prediction power of different machine-learning models before and after adding these features to the state-of-art requested permissions features in Android. We used four popular ML classifiers (i.e. Random Forest, MultiLayer Perceptron neural networks, Decision trees, and Naïve Bayes) for our experiments and we found that the new set of features have a tiny improvement in the performance in the case of decision trees and Random forest classifiers
- 4) Mobile Botnet Detection: A Deep Learning Approach Using Convolutions Neural Networks:- Android, being the most widespread mobile operating system is increasingly becoming a target for malware. Malicious apps designed to turn mobile devices into bots that may form part of a larger botnet have become quite common, thus posing a serious threat. This calls for more effective methods to detect botnets on the Android platform. Hence, in this paper, we present a deep learning approach for Android botnet detection based on Convolutional Neural Networks (CNN). Our proposed botnet detection system is implemented as a CNN- based model that is trained on 342 static app features to distinguish between botnet apps and normal apps. The trained botnet detection model was evaluated on a set of 6,802 real applications containing 1,929 botnets from the publicly available ISCX botnet dataset. The results show that our CNN-based approach had the highest overall prediction accuracy com- pared to other popular machine learning classifiers. Furthermore, the performance results observed from our model were better than those reported in previous studies on machine learning-based Android botnet detection
- 5) Detection of Mobile Botnets using Neural Networks:- This poster deals with botnets, the most dangerous kind of mobile malware, and their detection using neural networks. Unlike common mobile malware, botnets often have a complicated pattern of behavior because they are not managed by predictable algorithms but they are controlled by humans via command and control servers (CC servers) or via peer-topper networks. However, they have certain common features which have been revealed by analysis of contemporary mobile botnets.

These features have been used for the creation of a neural network training set. Finally, the design of parallel architecture using neural network for useful detection of mobile botnets has been described

3. PROPOSED SYSTEM

3.1. System Architecture

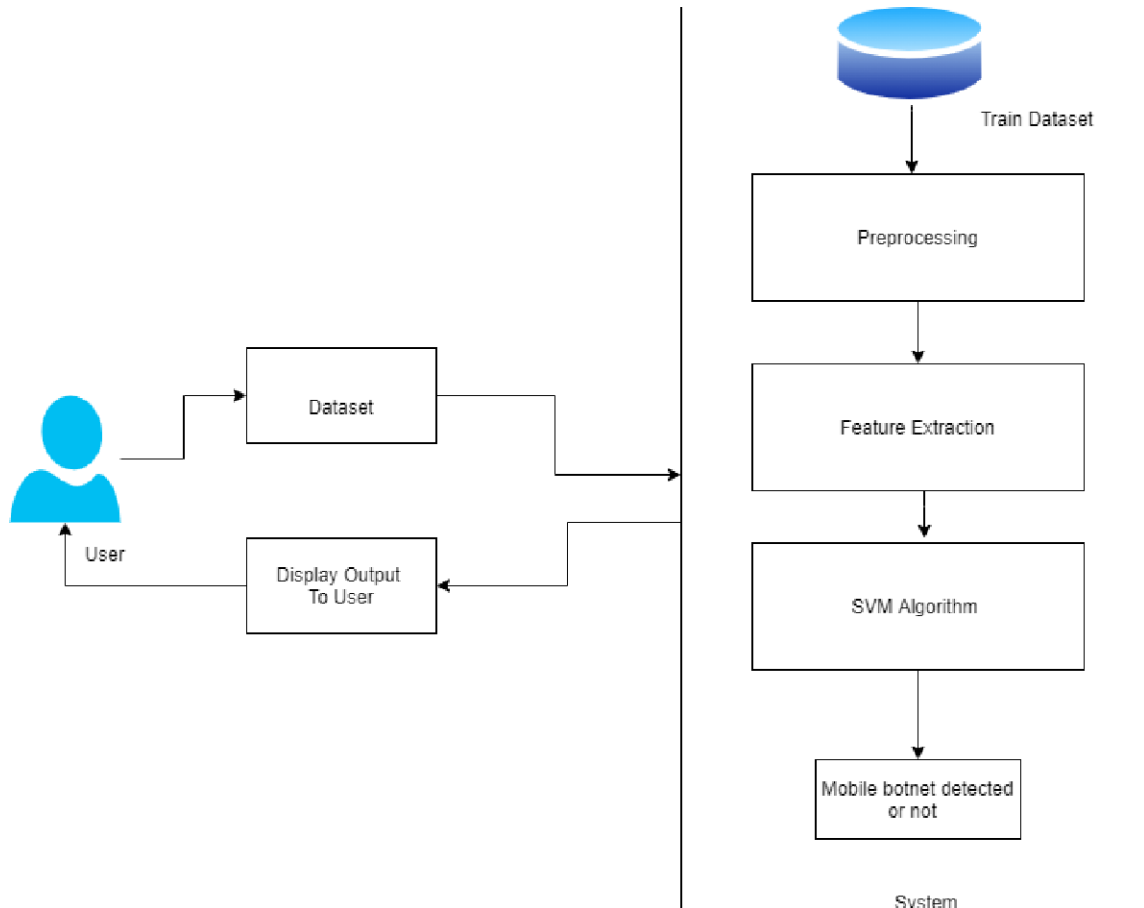


Fig. 1. System Architecture of Mobile Botnet Detection

3.2. data flow diagram

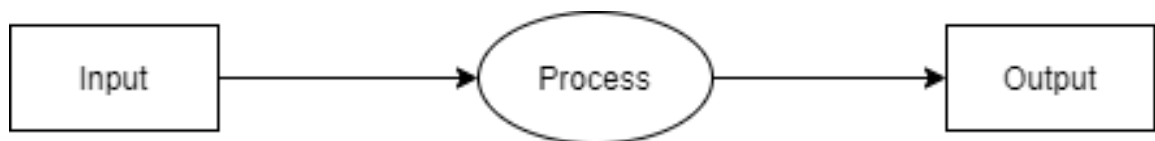


Fig. 2. Data flow diagram 0

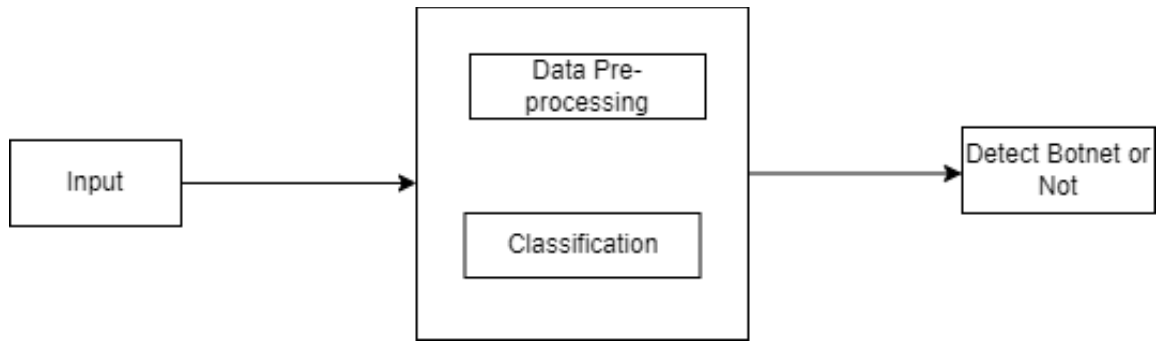


Fig. 3. Data flow diagram 1

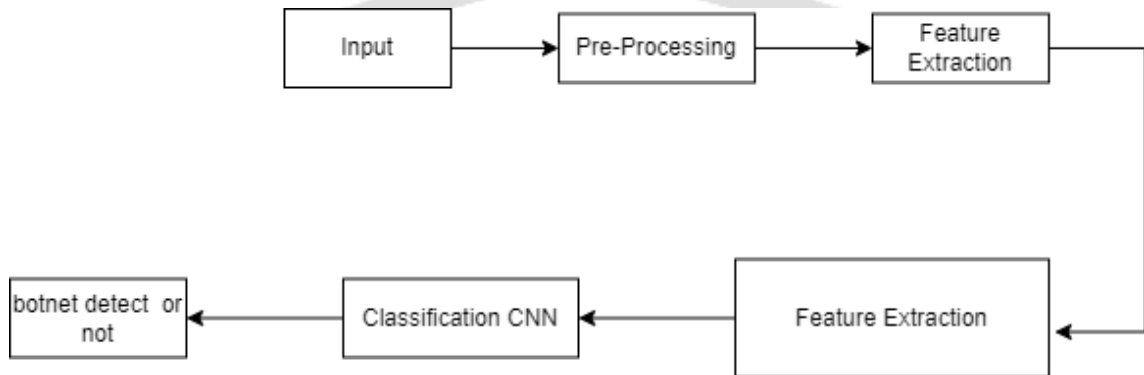


Fig. 4. Data flow diagram 2

3.3 class diagram

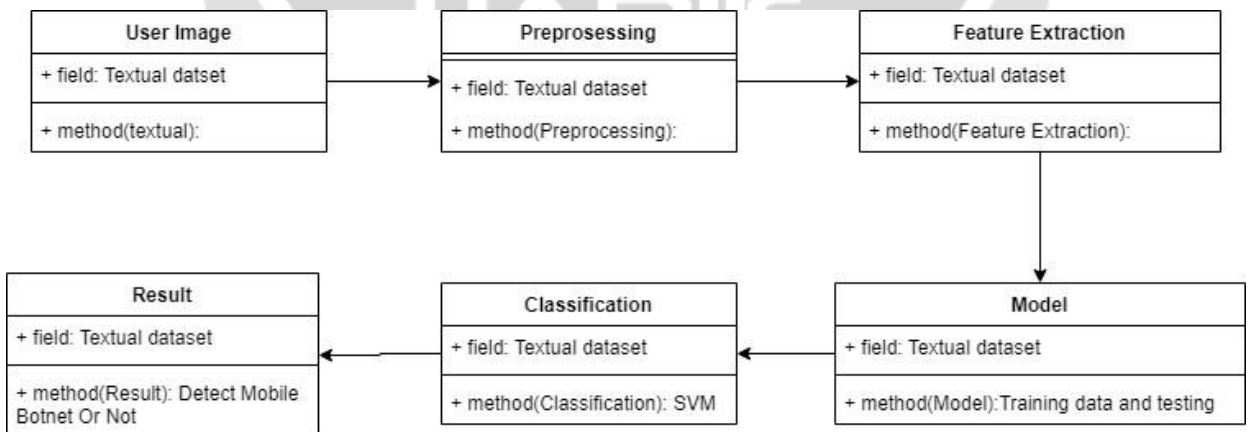


Fig.5. Class diagram

3.4 Use case diagram

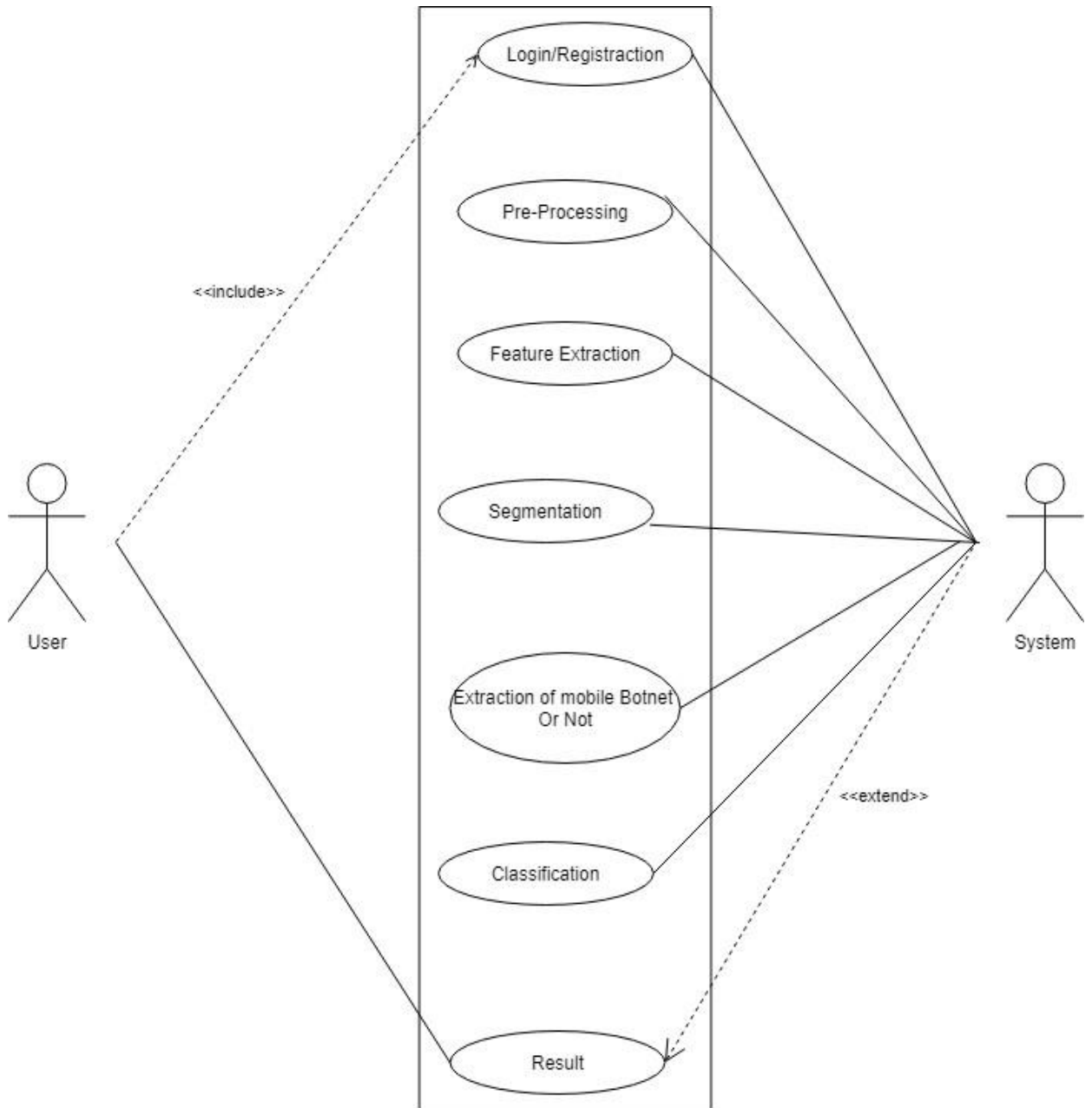


Fig.6. Use case diagram

4. RESULT

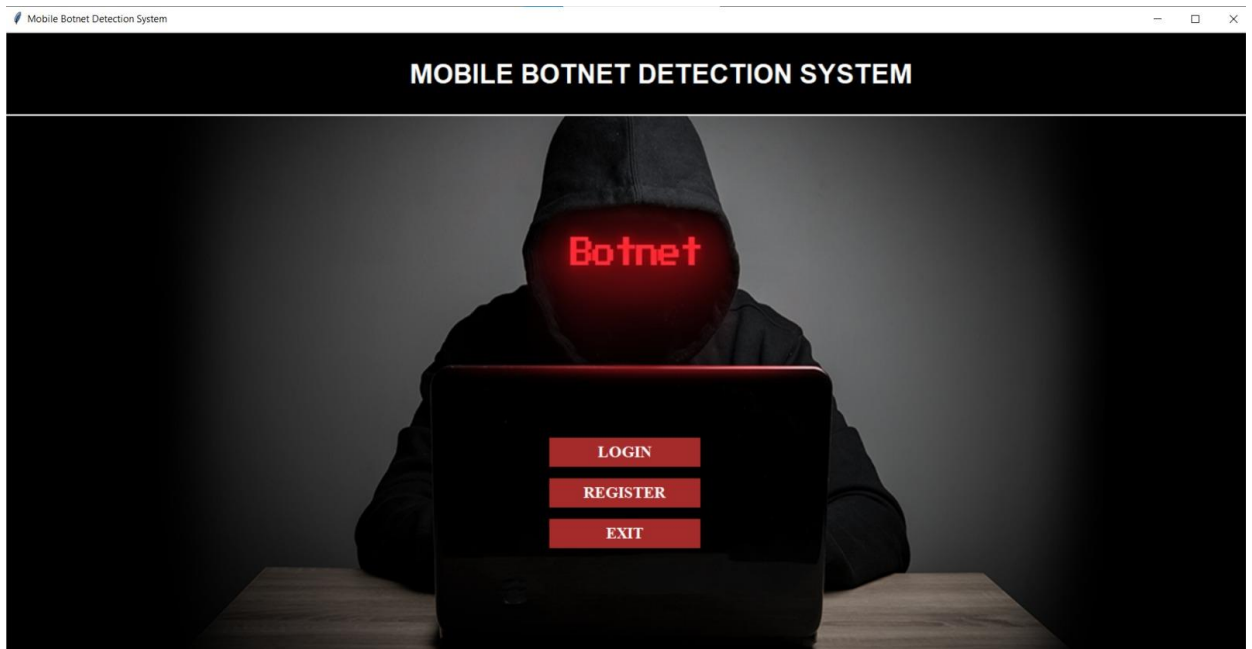


Fig.1. Home Page

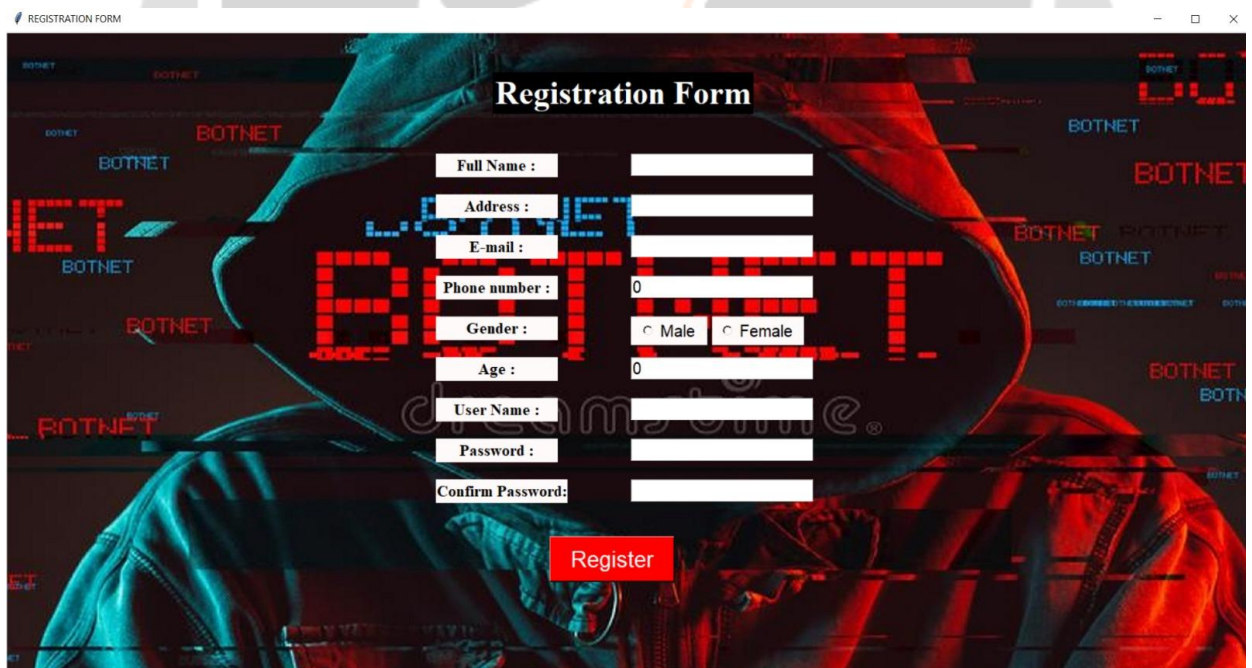


Fig.2. Registration

page

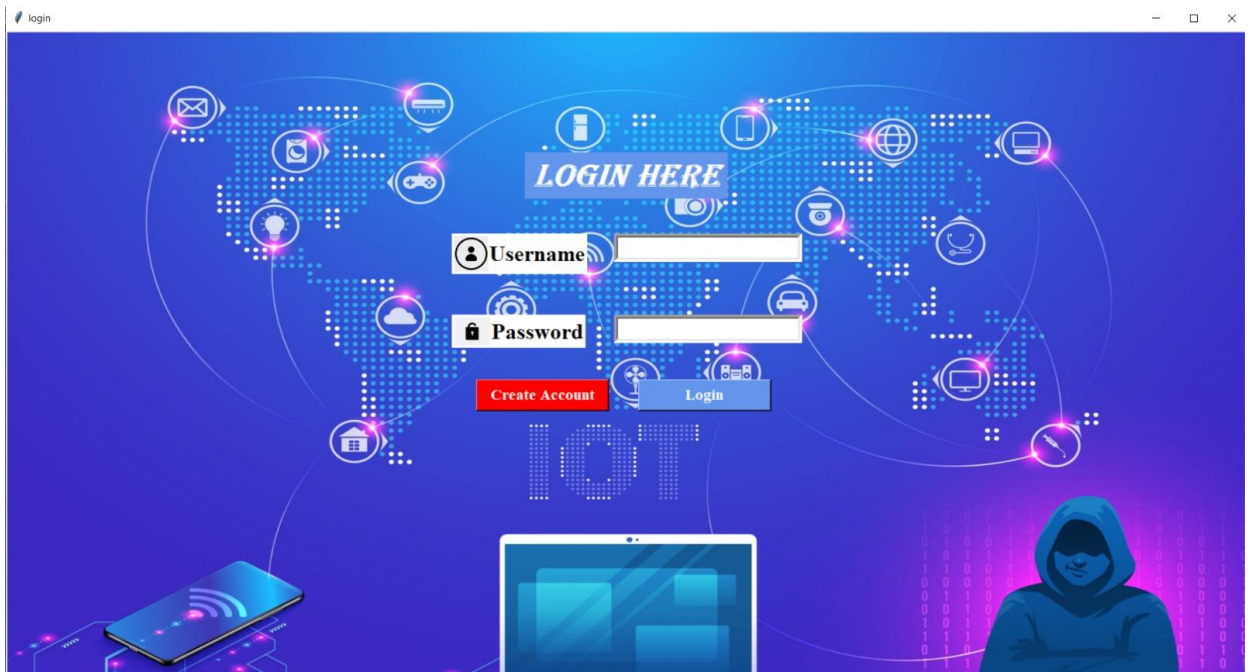


Fig.3 Login Page



Fig.4. Botnet detection home page

Telephony Get Device ID	0	Chmod	0
Telephony Get Subscriber ID	0	Mount	0
Abort Broadcast	0	.APK	0
Send SMS	0	.ZIP	0
Delete Packages	0	.DEX	0
Phone_State	0	install_packages	0
SMS_Received	0	Android.intent.action.Battery_low	0
Ljava.net.InetSocketAddress	0	.SO	0
Read_SMS	0	Android.intent.action.power_connected	0
Android.intent.action_Boot_completed	0	System.* Load Library	0
IO.File.*delete	0	.EXE	0
Chown	0		

Submit

Fig.5. Botnet detection

page

How to prevent

- Keep all systems updated
- Provide user awareness training
- Multi-factor authentication (MFA)
- Monitor network traffic
- Adopt a passwordless environment
- Implement zero trust
- Exit

Botnets are designed to exploit vulnerabilities in your network, which includes unpatched security risks in connected devices. Keep those devices more secure by installing antivirus and other software updates and patches as soon as they become available. Even if they're not actively used, all hardware and legacy devices should be kept up to date.

Fig.6. Botnet prevention

page

5. ADVANTAGES

- Mobile botnets take advantage of unpatched exploits to provide hackers with root permissions over the compromised mobile device, enabling hackers to send email or text messages, make phone calls, access contacts and photos, and more
- Provide more security
- Easy to handle

6. USE OF APPLICATION

- Company
- In Office
- In Banking
- All User

7. CONCLUSION

Botnets are a Dangerous evolution in the malware world. They are being used to damage systems, steal information, and Comprise Systems. They are hard to detect and eliminate. so our system is useful to Detect Mobile Botnet

8 . FUTURE WORK

With just one neural network, it is feasible to recognize features from both sets. To estimate the natural ratio of these sets, a static analysis of both sets with a significant number of samples is required (such that the central limit theorem is valid). The training set will be constructed in such a way that vectors with this ratio are included.

9. REFERENCES

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