



## **Financial Botnets - Online Threat for Financial Banking**

**M.D. Amala Dhaya,**  
Assistant Professor in Computer Science Department  
Loyola Institute of Technology of Science

**Dr. R. Ravi**  
Professor & Research Centre Head,  
Department of Computer Science and Engineering  
Francis Xavier Engineering College,  
Tirunelveli - 627003, Tamil Nadu State, India.  
[fxhodcse@gmail.com](mailto:fxhodcse@gmail.com)

### **ABSTRACT**

New way of financial flow through Electronic links motivates cyber criminals to keep their botnets modest, to target financial networks, smaller banks and compromise other types of networks for financial gains. Cybercriminals have proven themselves to be a resilient bunch. It seems like for every major takedown of a botnet or strain of malware, there are hackers already poised to make the next move in the chess match between cybercriminals and the security intelligence community. The proposed paper focuses on Threats for Online Financial Banking on Financial Botnets.

**Keyword-** Botnets, anomaly, phishing websites

### **1. INTRODUCTION**

On each successive day, thousands of computers in commercial environment are infected with more number of financial malware like, Zeus that enable them to be a part or a to become zombies, which are able to bond huge financial botnets that can be take on by efficient cyber-criminals in order to whip online banking clients identification in a financial network. In spite of the fact that finding and easing mechanisms for spam and DDoS associated botnets have been extensively researched and build up. It is true that the inactive nature (i.e. low network traffic, less connections) of financial botnets very much delay their process to prevent, or mitigate the effects of, threats to an entity.

As a result, cyber-criminals are still making high financial earnings at comparatively small risk with the help of financial botnets. In this text we suggest the use of openly available IP blacklists to



sense both drones and C&C nodes that are element of financial botnets, based on their behaviour. To show this hypothesis we have developed an official framework capable of assessing the eminence of a blacklist, by compare it, in opposition to a baseline and taking to consider dissimilar metrics [1-3].

The contributed structure has been tested with about 500 million IP addresses, gathered during one-month phase from seven dissimilar famous blacklist donors. Our experimental outputs showed that, this blacklisted IPs are intelligent to sense both drones and C&C connected with the Zeus botnet and very important, that it is potential to assign dissimilar excellence scores to every blacklist based on our assessments. Finally, we bring in the essentials of a high-performance IP standing system that use the earlier collected blacklists excellence scores, in order to respond, almost in real-time scenario whether a sure IP is a member of a financial botnet or not. Our principle is that such an arrangement can be easily included into e-banking anti-fraud structures [4-7].

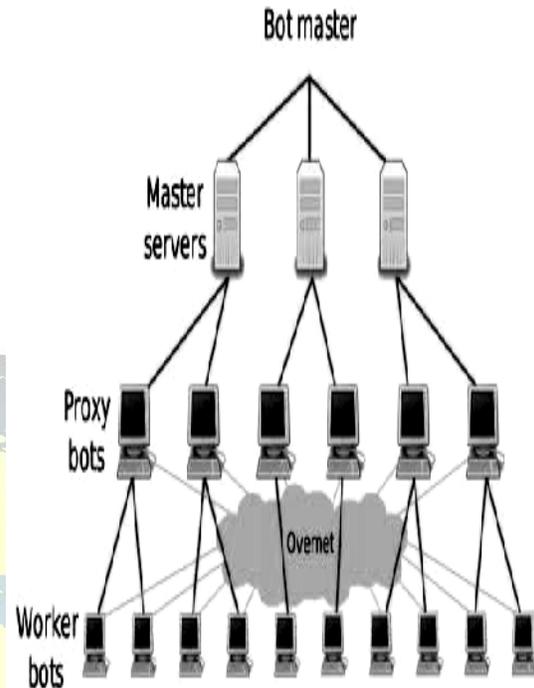


Figure 1 Botnet Architecture

Figure 1 shows the general Botnet Architecture of the proposed scheme. Attacking Behaviors are Distributed Denial-of-Service Attacks, Spamming, Sniffing Traffic, Key logging, Spreading new malware, Installing Advertisement Addons and Google AdSense abuse.

### 3. RESULT AND DISCUSSION

## 2. PROPOSED SYSTEM

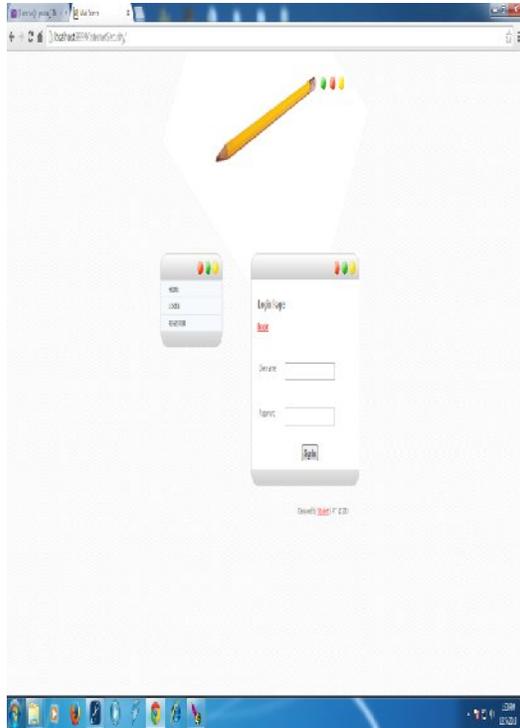


Figure 2 Email Login Page

Figure 2 shows the Email Login Page of the proposed botnet.

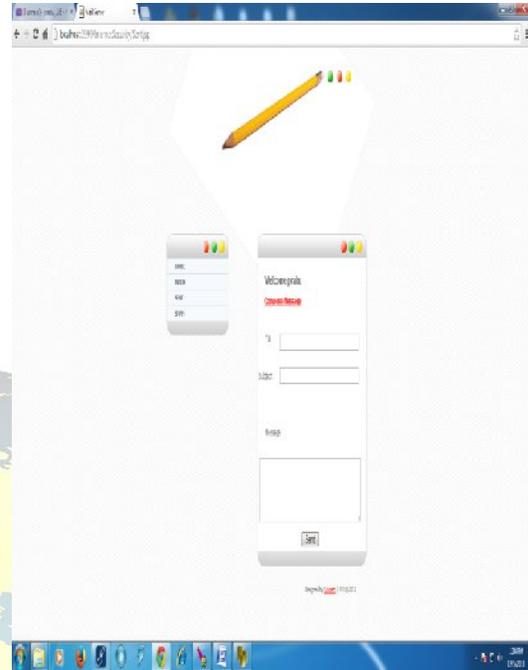


Figure 3 Email Received in Spam

Figure 3 shows the Email Received in Spam of the proposed botnet.

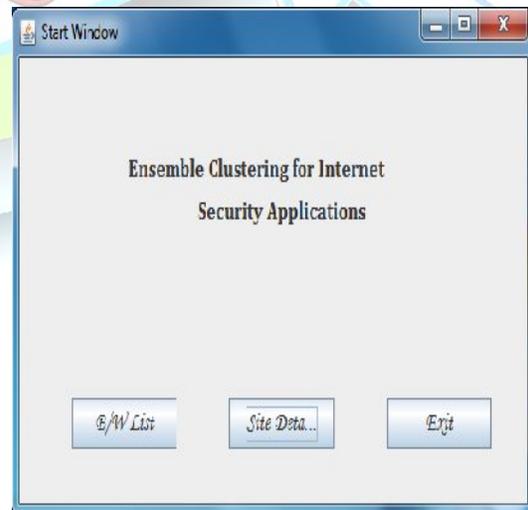




Figure 4. Starting page

Figure 4 shows the Starting page of the proposed botnet.

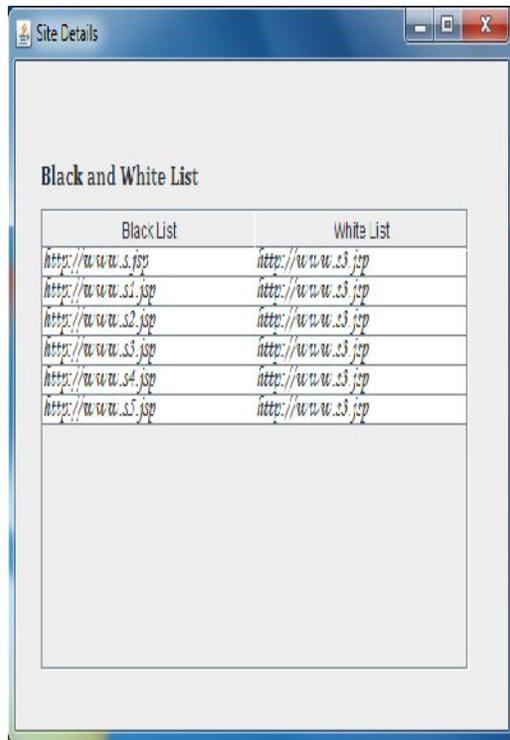


Figure 5 Black list and white list

Figure 5 shows the Black list and white list of the proposed botnet.

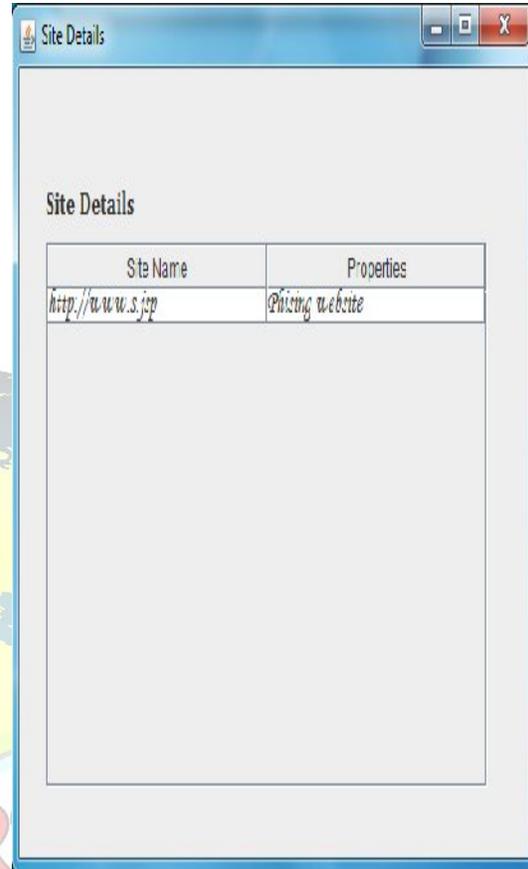


Figure 6 Site Details

Figure 6 shows the Site Details of the proposed system .

#### 4. CONCLUSION

The financial fraud marketplace is a progressively more prearranged unit. It is a service-based business in which a wide variety of financial webinjects, trojans, and malwares and allocation channels are bought and sold. Invaders are also attaining new markets, continually getting



bigger their operations to places, where they can relate existing technique. Thus the Threats for Online Financial Banking on Financial Botnets are routine attack detection requires gathering, joining, and repeatedly analyzing statistics to extract pertinent information and apply protection countermeasures. Joining this data with information collected on known botnets will help expand the knowledgebase for identifying attacks and selecting suitable attack alleviation tools.

## 5. REFERENCES

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