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# Detection of Phishing Websites using Machine Learning

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Abstract: There are number of clients who purchase products online and make payment through various websites and also there are multiple websites who ask clients to provide sensitive data such as username, password or credit card details etc. often for malicious reasons. This type of websites is familiar as phishing website. So, to disclose and predict phishing website, we suggested a quick, flexible and effective system that is based on classification Random forest algorithm. We utilize the Random forest algorithm and techniques to extract the phishing data sets criteria to classify their legitimacy. The phishing website can be detected based on some important characteristics i.e URL, Domain Identity, Security and encryption criteria in the final phishing detection rate. Once user makes transaction through online when he makes payment through the website our system will use Random forest algorithm to detect whether the website is phishing website or not. This application can be utilized by many E-commerce enterprises in order to make the whole transaction process secure. Random forest algorithm used in this system provides better performance as compared to other traditional classifications algorithms, with the help of this system user can also purchase products online without any hesitation. Administrant can add fraud website URL into system where system could access and scan the fraud website and by using algorithm, it will add new suspicious keywords to database. System make use of machine learning technique to add new keywords into database.

Keywords: Random forest algorithm, machine learning, classification.

INTRODUCTION- In today's world, technology has become an integral part of the 21st century. The Internet is one of this technology, which is growing rapidly every year and plays an important role in people's lives. It's already an important and easy way to support social transactions such as e-banking and e-commerce transactions. That has led users to believe that it is easier to disclose their personal information on the Internet. As a result, security thieves have begun to identify this information as a major security problem.

Crime information theft websites are considered one of these problems. They use a social engineering strategy, viz can be described as fraudsters trying to trick a user into giving them their personal information exploiting human weaknesses rather than software

To understand the impact of sensitive identity theft, we need to be aware of the different types of sensitive identity theft such as,

### Fraudulent theft of sensitive information:

Identity theft is the most common type of identity theft. In this case, the attacker is trying to find the secret information from victims. Attackers use the information to steal money or to make other attacks.

# **Spear Phishing:**

Spear Phishing identifies certain people instead of a broad group of people. Attackers often reach out to their victims on social media and other sites. That way, they can customize their communication and make it appear more realistic. Smell When the attackers follow the "big fish" as CEO, it's called Whaling. These attackers often spend a lot of time to enter the target information to find the correct time and methods to steal the login information Similar to stealing sensitive information, Pharming sends users to a fraudulent website that appears to be legitimate. However, in this case victims don't even have to click a bad link to be redirected to a fake site. Theft of sensitive information is one of the biggest problems for information security. It can happen in two ways, it can be by finding suspicious emails lead to a fraudulent site or by users accessing links directly to the website of a phishing scam. However, these two approaches are common to one thing, namely that the attacker looks at human weakness rather than there is a software crash. Identity theft can be defined as fraudulent attempts to trick a user into giving out personal information, such as a username, password, and credit card number. These scams lead in economic and financial problems for users.

A URL with a database that already contains a list of URLs of a website for phishing scams. Due to the rapid growth of websites that steal sensitive information, the terrorist method does not work well in determining that each URL is a phishing scam website or not, and this type of delay can lead to zero- day attacks from new sites to steal sensitive information. Machine learning benefits from its ability to predict. Learns the features of a website to steal sensitive information as well and predicts new aspects of identity theft. There are many technical options, such as the Nave Bayes (NB), the decision tree (DT), vector support equipment (SVM), RF, neural input network (ANN), and Bayesian net (BN). Accuracy detection of identity theft varies from one algorithm to another.

In this work we used a clever forest algorithm and techniques to pull out the process of setting up crime information theft sensitive information to distinguish their legitimacy. A website for the theft of sensitive information can be accessed according to certain important factors namely URL, domain ownership, security and encryption methods are ultimately detected for phishing scams,

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#### LITERATURE SURVEY

Fraudulent identity theft is a fraudulent attempt to obtain sensitive personal or organizational information usernames, passwords and credit card details by transforming them into reliable electronic devices communication. Attacks on sensitive identity theft pose significant threats to the privacy and security of the user.

The purpose of this study is to present an overview of the theft of various sensitive information and various strategies. data protection. Includes a discussion of Extreme Learning Machine (ELM) based 30 feature classification including phishing website data in UC Irvine Machine Learning repository details.

A cybercrime website has emerged as a major cyber security threat in recent times. Crime websites for stealing sensitive information spam management, malware, malware, driver- driven, etc. A criminal website for stealing sensitive information is often similar to a very popular and attractive website for an unexpected user to fall into the trap of. The victim of fraud causes financial loss, loss of confidential information and loss of dignity. Therefore, it is important to find a solution that can reduce those security threats in a timely manner. Traditionally, the detection of criminal websites for the theft of sensitive information is done through a blank list. There are many popular websites that have a list of crude websites, e. g. Phis Tank. Prohibition technology is lacking in two things, the black list may be incomplete and may not find the latest innovation details of criminal theft of sensitive information Website theft of sensitive information is one of the most common cybersecurity problems Software vulnerability. It can be described as a process of attracting online users to their own sensitive information such as usernames and passwords.

Cybercrime is an easy way to get sensitive information from innocent users. The purpose of the fishermen for sensitive information such as username, password and bank account details. The cyber security people are the ones now i am looking for reliable and solid strategies for finding phishing websites.

This paper collaborates with machine learning technology to detect criminal URL theft URLs through various extracts and analyzes features of official URLs and criminal identity theft. Tree decision, random forest and Support vector machine algorithms used to detect phishing websites. The purpose of this paper is to find criminal URLs for stealing sensitive and sensitive information significantly reduced algorithm learning by comparing accuracy, false and false measurement each algorithm.

additional functionality in the Internet browser as an extension that automatically notifies the user there finds a website to steal sensitive information. The program is based on a machine learning method, especially monitored to read.

#### PROPOSED SYSTEM:

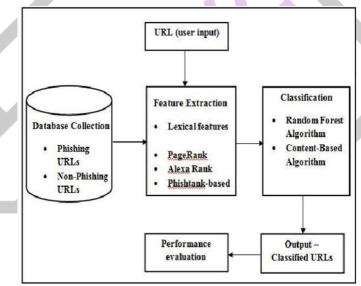


Figure 1. Block Diagram of phishing website detection

The above figure shows how a sensitive identity theft website will be located when the user has to enter any website URL that contains the subdomain and subdomain, the method used is to extract from the lexical URL attributes, page rank, Alexa Rank etc. then randomly works a forest algorithm that will first prioritize these features where features already exist in the database which will be called training data set randomly of the forest algorithm to predict whether the website is a phishing scam or a standard existing database.

## **Feature Extraction**

Ranking.Page: PageRank works by calculating the number and quality of page links to determine the critical limitations of how important a website is. The value assigned to a web page as a measure of its popularity or value, is used to determine the order of the search engine results presented.

ALEXA Rank: Commonly used by online businesses for competitive analysis. Alexa ranks are a link between people who are estimated to visit the site and how many pages are viewed.

Phish Tank: Phish tank is a public crime verification system where users file suspected threats, and other users in the system vote to see if threats of identity theft are legitimate

#### **Random Forest Algorithm**

Random Forest is a popular algorithm for machine learning that is a supervised learning method. It can be used for both Separation ML problems. It is based on the concept of learning together, which is the process of combining multiple divisions to solve a complex problem and improve the performance of a model. Since the unplanned forest includes many trees to predict the database phase, it is possible that some decision trees may predict correct extraction, while others may. But collectively, all trees predict the right yield. Therefore, below are two aspects of the best random segregation of the forest: There should be some real value for the data feature variables so that the separator can predict accurate results rather than an estimated result. Predictions from each tree should have very low interactions. Therefore, this database is provided by a random forest divider. The database is divided into subsets and is assigned to each decision tree.

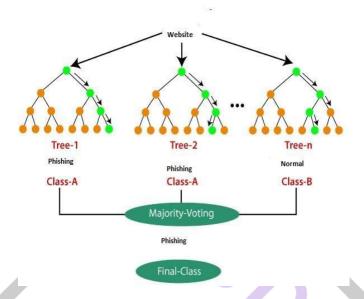


Figure 2.Tree diagram

#### **IMPLEMENTATION:**

We have many modules like Dashboard, login page, user view, add to blacklist, add names, watch list, view feedback, check website, we have used two modules namely Dashboard and login page

# Modules and their Description

The system comprises of 6 major modules as follows:

- 1. Registration
- 2. Login
- 3. Add to Blacklist
- 4. Check Website
- 5. Feedback
- 6. Change Password

Registration: A visitor can register himself to the website to access it.

Login: After a successful registration, user/admin may input his credentials to login into the system.

**Add to Blacklist**: Here, the system administrator adds the malicious website to the blacklist. **Check Website:** Here, the user checks for the blacklisted website by inputting the URL.

Feedback: A user could send a feedback regarding the website to the admin.

Change Password: Admin may change his password for security purpose by inputting old and new password.

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# **RESULTS:**

# **Phishing Website**





Figure 3.Login Page

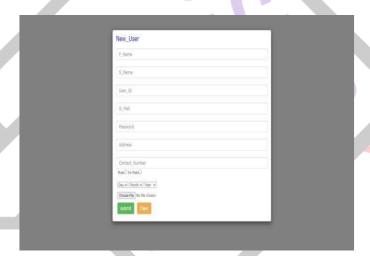


Figure 4.Sign Up Page



Figure 5. Dashboard

#### CONCLUSION

Unfamiliar on phishing education makes the attack successful. Even with the help of few indicators used by the browser such as pad lock identification, lock icon, and site identity button, the user still cannot identify the attack. Web spoofing attack is difficult to detect. Even with the current security prevention method, these attacks still occur. The main moto of This study is to help the users specially to differentiate between the legitimate and phishing web pages by using url as an indicator. Finding of this Research demonstrates its ability to identify the fake webpages based on their urls. As a conclusion, the most important way to protect the user from phishing attack is the education awareness. Internet users must be knowledgeable of all security tips which are given by experts. In the phishing websites dataset, 4898 phishing websites and 6157 legitimate websites were gathered and used for training and evaluating the supervised machine Learning classifiers used in phishing websites detection.

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