WEB PHISHING DETECTION

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1.INTRODUCTION

1.1 PROJECT OVERVIEW

The criminals, who want to obtain sensitive data, first create unauthorized replicas of a real website and e-mail. The e-mail will be created using logos and slogans of a legitimate company. The nature of website creation is one of the reasons that the Internet has grown so rapidly as a communication medium. Phisher then send the "spoofed" e-mails to as many people as possible in an attempt to lure them into the scheme. When these e-mails are opened or when a link in the mail is clicked, the consumers are redirected to a spoofed website, appearing to be from the legitimate entity. We discuss the methods used for detection of phishing Web sites based on URL importance properties.

Phishing has been accounted for many fraudulent incidents on the internet in the recent years, and it is showing no sign of stopping anytime soon. So, what is phishing? It is a term that is used to describe a malicious individual or a group of individuals who scam users. This is done by sending emails or creating web pages that are designed to collect an individual's online credentials, credit card details or other login information's. The concept of detecting phishing websites is usually done by looking through a huge database or a directory that contains all the malicious sites that has been logged by internet users or community members. An effective way for end users to benefit from phishing detection is by having the option to use an extension plugin that works on real time, as it gives them real time indication of what they are surfing and as well as if they are safe while browsing.

1.2 PROJECT PURPOSE

The main purpose of the project is to detect the fake or phishing websites who are trying to get access to the sensitive data or by creating the fake websites and trying to get access of the user personal credentials. We are using machine learning algorithms to safeguard the sensitive data and to detect the phishing websites who are trying to gain access on sensitive data.

This research mainly will focus on implementing machine learning in JavaScript for it to run on a browser as an extension since JavaScript does not have much library support towards Machine Learning and also to keep in mind of the users' machines performance. This approach should be made with the intention of having it lite in order to achieve the capability to allow as much users as possible to use it.

Random forest classifier for this project will be trained traditionally based on the phishing dataset 2 using Python scikit, and parameters of this model will then be exported in a JSON format to be used together with JavaScript.

Phishing is a form of fraud in which the attacker tries to learn sensitive information such as login credentials or account information by sending as a reputable entity or person in email or other communication channels.

Typically a victim receives a message that appears to have been sent by a known contact or organization. The message contains malicious software targeting the user's computer or has links to direct victims to malicious websites in order to trick them into divulging personal and financial information, such as passwords, account IDs or credit card details.

The main reason is the lack of awareness of users. But security defenders must take precautions to prevent users from confronting these harmful sites. Preventing these huge costs can start with making people conscious in addition to building strong security mechanisms which are able to detect and prevent phishing domains from reaching the user

2. LITERATURE SURVEY

2.1 Existing Problem

Cyber criminals use phishing emails because it's easy, cheap and effective. Email addresses are easy to obtain, and emails are virtually free to send. With little effort and cost, attackers can quickly gain access to valuable data. Those who fall for phishing scams may end up with malware infections (including ransomware), identity theft and data loss.

The data that cybercriminals go after includes <u>personal identifiable</u> <u>information (PII)</u>—like financial account data, credit card numbers and tax and medical records—as well as sensitive business data, such as customer names and contact information, proprietary product secrets and confidential communications. Cybercriminals also use phishing attacks to gain direct access to email, social media and other accounts or to obtain permissions to modify and compromise connected systems, like point-of-sale terminals and order processing systems.

Many of the biggest data breaches, like the headline-grabbing 2013 Target breach, start with a phishing email. By using a seemingly innocent email, cybercriminals can gain a small foothold and build on it. in order to begin the development of a google chrome browser extension, it should emphasize on the ability to alert and warn the users if they accidentally visited a phishing webpage. This chrome extension will also be developed keeping in mind that, it should not have any 3rd party servers or API present to call services as this gives a narrow path for hackers to target users browsing pattern. Lastly, this extension plugin will also provide an instantaneous detection service that warns users as they view a phishing website, just so they avoid entering any confidential information before it is too late.

2.2 References

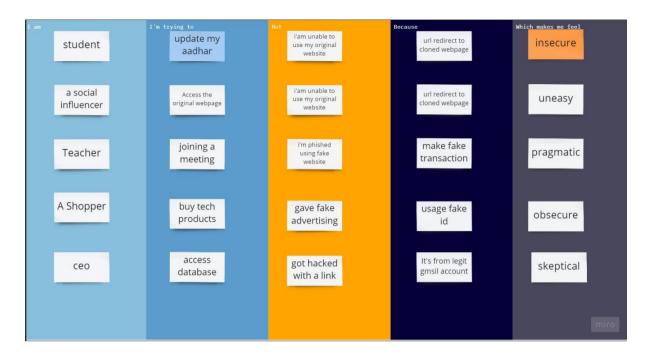
Phishing is a form of fraud in which the attacker tries to learn sensitive information such as login credentials or account information by sending as a reputable entity or person in email or other communication channels.

Typically a victim receives a message that appears to have been sent by a known contact or organization. The message contains malicious software targeting the user's computer or has links to direct victims to malicious websites in order to trick them into divulging personal and financial information, such as passwords, account IDs or credit card details.

Phishing is popular among attackers, since it is easier to trick someone into clicking a malicious link which seems legitimate than trying to break through a computer's defense systems. The malicious links within the body of the message are designed to make it appear that they go to the spoofed organization using that organization's logos and other legitimate contents. In this article I explain: phishing domain (or Fraudulent Domain) characteristics, the features that distinguish them from legitimate domains, why it is important to detect these domains, and how they can be detected using machine learning and natural language processing techniques.

By: Ebubekir Bubar, Ajay, Christian

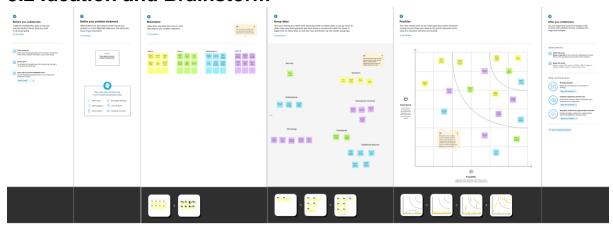
2.3 Problem Statement Definition



3.IDEATION & PROPOSED SOLUTION 3.1 EMPATHY MAP CANVAS



3.2 Ideation and Brainstorm



3.3 Proposed Solution

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	Many banking websites, and the anonymous links for malicious content, request users to submit private information like usernames, passwords, credit card numbers, and more. Phishing websites are this kind of online banking website. One of the many security risks to web services on the Internet is web phishing.
2.	Idea / Solution description	1. Address Bar based Features 2. Abnormal Based Features 3. HTML and JavaScript Based Features 4. Domain Based Feature. are the components are used for detection of phishing attack
3.	Novelty / Uniqueness	The proposed solution uses novel methods to identify relevant features by factoring in attributes like page rank, in-degree, presence of HTTPS token, and others. Through the features obtained, different dataset sizes and distributions are tested for analyzing and detecting optimal hyperparameters.
4.	Social Impact / Customer Satisfaction	The website is created with an opinion such that people are not only able to distinguish between legitimate and fraudulent website, but also become aware of the malpractices. They can stay away from the people trying to exploit one's personal information, like email address, password, debit card numbers, credit card details, CVV, bank account numbers, and the list goes on.
5.	Business Model (Revenue Model)	Micro web frameworks like flask can be used to create a REST-based

		web application and web services that users may use to conduct reliable and secure online transactions through safe ecommerce websites and identify the illegal links. Based on membership levels, different levels of security strictness and multiple volumes of secure ecommerce websites would be offered.
6.	Scalability of the Solution	The proposed website's functionality can be turned into an API so that other e-banking and e-commerce portals can utilize it to provide their users with safer and more effective solutions. It could be further extended for other security concerns such as audio recording, video recording, location tracking, virus attacks, and more with safer and more effective solutions.

3.3 Problem Solution Fit



4.REQUIREMENT ANALYSIS:

4.1 Functional Requirement

FR NO.	Functional	Sub Requirement (Story / Sub-Task)
	Requirement	
	(Epic)	
FR-1	User Input	User inputs an URL in required field to check
		its validation.
FR-2	Website	Model compares the websites using Blacklist
	Comparison	and Whitelist approach.
FR-3	Feature extraction	After comparing, if none found on comparison
		then it extracts feature using heuristic and
		visual similarity approach.
FR-4	Prediction	Model predicts the URL using Machine
		Learning algorithms such as Logistic
		Regression, KNN
FR-5	Classifier	Model sends all output to classifier and
		produces final result.
FR-6	Announcement	Model then displays whether website is a legal
		site or a phishing site.
FR-7	Events	This model needs the capability of retrieving
		and displaying accurate result for a website

4.2 . Non-functional Requirements:

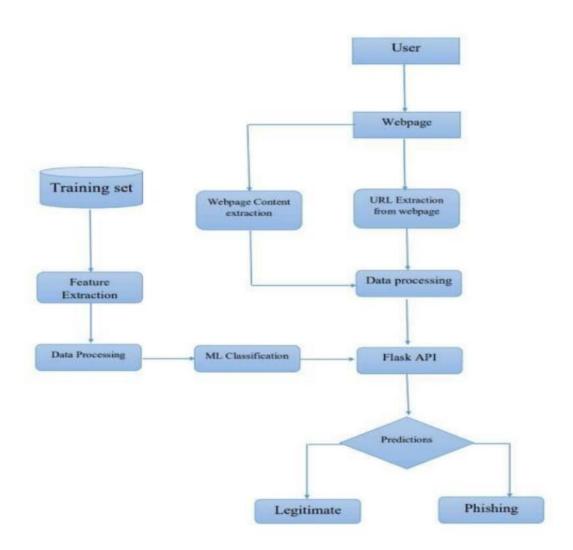
Following are the non-functional requirements of the proposed solution.

FR No.	Non-Functional Requirement	Description
NFR-1	Usability	Any URL must be accepted for detection
NFR-2	Security	Alert message must be send to the users to enable secure browsing.
NFR-3	Reliability	The Phishing websites must detected accurately and the results must be reliable
NFR-4	Performance	The performance and interface must be user friendly
NFR-5	Availability	Anyone must be able to register and login
NFR-6	Scalability	It must be able to handle increase in the number of users.

5. PROJECT DESIGN

5.1 Data Flow Diagrams

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.

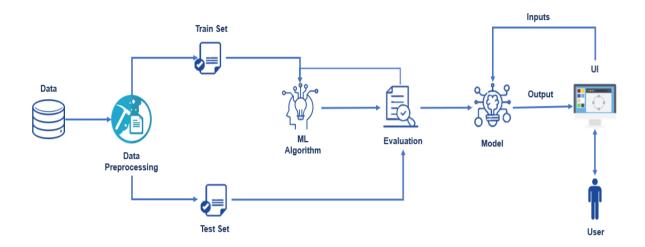


Use the below template to list all the user stories for the product.

User Type	Functional Requireme nt (Epic)	User Story Numbe r	User Story / Task	Acceptance criteria	Priority	Release
Custom er (Mobile user)	Registratio n	USN-1	As a user, I can register for the application by entering my email, password, and confirming password.	I can access my account / dashboard	High	Sprint-1
		USN-2	As a user, I will receive confirmation email once I have registered for the application	I can receive confirmation email & click confirm	High	Sprint-1
		USN-3	As a user, I can register for the application via Facebook		Low	Sprint-2
		USN-4	As a user, I can register for the application via Gmail		Medium	Sprint-1
	Login	USN-5	As a user, I can log into the application by entering email & password		High	Sprint-1
	Dashboard					
Custom er (Web user)	User input	USN-1	As a user i can input the particular	I can go access the website	High	Sprint-1

			URL in the required field and waiting for validation.	without any problem		
Custom er Care Executiv e	Feature extraction	USN-1	After i compare in case if none found on comparison then we can extract feature using heuristic and visual similarity approach.	As a User i have comparison between websites for security.	High	Sprint-1
Adminis trator	Prediction	USN-1	Here the Model will predict the URL websites using Machine Learning	In this i can have correct prediction on the particular algorithms	High	Sprint-1

5.2 Solution & Technical Architecture



5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptation Criteria	Priority	Release
Customer (web user)	Website	USN – 1	Some, times the user need to check the website to check particular URL is safe or not	Website has easy to use and responsive.	High	Sprint-1
	Alter Notification	USN – 2	If enter into some malicious link, notification has to be sent to me	Received notification in mobile or to my mail ID	Low	Sprint-1
	Blocking	USN – 3	Whenever the link is not safe to enter, it should block to use to me.	I can register and access the dashboard with Facebook login.		Sprint-2
	Allowing	USN – 4	If I wish to use that website then it should also allow me to enter into that website			Sprint-1
	Login	USN - 5	As a user, I can log into the application by entering email & password	The phishing website has to be determined correctly.	High	Sprint-1

Customer (web view)	User input	USN - 1	As a user I can enter the required URL in the box while awaiting validation	I can access the website without any problem	High	
Customer care executive	Feature extraction	USN - 1	In the event that nothing is discovered during comparison, we can extract features	As a user I can have comparison between websites for security	High	
			using a heuristic and a visual similarity technique.			
administrator	Prediction	USN - 1	The model will use machine learning algorithms like a logistic regression and KNN model to forecast the URL of the websites.	I can accurately forecast the specific algorithms in this way	High	
	classifier	USN - 2	To create the final product. I will now feed all of the model output to classifier.	I will use this to identify the appropriate classifier for generating the outcome	Medium	Sprint-2

6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning And Estimation

TITLE	DESCRIPTION	DATE
Literature Survey & Information Gathering	Literature survey on the selected project & gathering information by referring the, technical papers, research publications etc.	16 OCTOBER 2022
Prepare Empathy Map	Prepare Empathy Map Canvas to capture the user Pains & Gains, Prepare list of problem statements	16 OCTOBER 2022
Ideation	List the by organizing the brainstorming session and prioritize the top 3 ideas based on the feasibility & importance.	16 OCTOBER 2022
Proposed Solution	Prepare the proposed solution document, which includes the novelty, feasibility of idea, business model, social impact, scalability of solution, etc.	16 OCTOBER 2022
Problem Solution Fit	Prepare problem - solution fit document.	16 OCTOBER 2022
Solution Architecture	Prepare solution architecture document.	16 OCTOBER 2022

Customer Journey	Prepare the customer journey maps to understand the user interactions & experiences with the application	16 OCTOBER 2022
Data Flow Diagrams	Draw the data flow diagrams and submit for review.	16 OCTOBER 2022
Technology Architecture	architecture diagram.	16 OCTOBER 2022

Prepare Milestone &	Prepare the	22 OCTOBER 2022
Activity	milestones & activity	
List	list of the project.	
Project Development -	Develop & submit the	24 OCTOBER 2022
Delivery of Sprint-1, 2,	developed code by	
3 & 4	testing it.	

7. CODING & SOLUTIONING (Explain the features added in the project along with code)

7.1 Feature 1

Detection:

Obtaining these types of features requires active scan to target domain. Page contents are processed for us to detect whether target domain is used for phishing or not. Some processed information about pages are given below.

- Page Title
- Meta Tags
- Text in the Body
- Images etc.,

By analysing these information formation, we can gather information such as:

- Is it required to login to Wesite
- Website category
- information about audience profile etc.,

All of features explained above are useful for phishing domain detection. In some cases, it may not be useful to use some of these, so there are some limitations for using these features. For example, it may not be logical to use some of the features such as ContentBased Features for the developing fast detection mechanism which is able to analyze the number of domains between 100.000 and 200.000. Another example would be, if we want to analyze new registered domains Page-Based Features is not very useful. Therefore, the features that will be used by the detection mechanism depends on the purpose of the detection mechanism. Which features to use in the detection mechanism should be selected carefully.

CODE:

<!DOCTYPE html>

```
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-
scale=1.0">
  <meta name="description" content="This website is develop for
identify the safety of url.">
  <meta name="keywords" content="phishing url,phishing,cyber
security, machine learning, classifier, python">
  <!-- BootStrap -->
  k rel="stylesheet"
href="https://stackpath.bootstrapcdn.com/bootstrap/4.5.0/css/bootstrap.
min.css"
    integrity="sha384-
9alt2nRpC12Uk9gS9baDl411NQApFmC26EwAOH8WgZl5MYYxFfc+Nc
Pb1dKGj7Sk" crossorigin="anonymous">
  <link href="static/styles.css" rel="stylesheet">
  <title>URL detection</title>
</head>
<body>
<div class=" container">
  <div class="row">
    <div class="form col-md" id="form1">
       <h2>PHISHING URL DETECTION</h2>
```

```
<br>
       <form action="/" method ="post">
          <input type="text" class="form input" name ='url' id="url"</pre>
placeholder="Enter URL" required="" />
          <label for="url" class="form label">URL</label>
          <button class="button" role="button" >Check here
        </form>
  </div>
  <div class="col-md" id="form2">
     <br>
     <h6 class = "right "><a href= {{ url }} target="_blank">{{ url
}}</a></h6>
     <br>
     <h3 id="prediction"></h3>
     <button class="button2" id="button2" role="button"
onclick="window.open('{{url}}')" target="_blank" >Still want to
Continue</button>
     <button class="button1" id="button1" role="button"</pre>
onclick="window.open('{{url}}')" target="_blank">Continue</button>
  </div>
</div>
<br>
</div>
  <!-- JavaScript -->
  <script src="https://code.jquery.com/jquery-3.5.1.slim.min.js"</pre>
```

```
integrity="sha384-
DfXdz2htPH0lsSSs5nCTpuj/zy4C+OGpamoFVy38MVBnE+lbbVYUew+
OrCXaRkfi"
    crossorigin="anonymous"></script>
  <script
src="https://cdn.jsdelivr.net/npm/popper.js@1.16.0/dist/umd/popper.min.j
s"
    integrity="sha384-
Q6E9RHvblyZFJoft+2mJbHaEWldlvl9IOYy5n3zV9zzTtml3UksdQRVvox
MfooAo"
    crossorigin="anonymous"></script>
  <script
src="https://stackpath.bootstrapcdn.com/bootstrap/4.5.0/js/bootstrap.min
.js"
    integrity="sha384-
OgVRvuATP1z7JjHLkuOU7Xw704+h835Lr+6QL9UvYjZE3Ipu6Tp75j7B
h/kR0JKI"
    crossorigin="anonymous"></script>
  <script>
```

```
let x = '{{xx}}';
let num = x*100;
if (0<=x && x<0.50){
    num = 100-num;
}
let txtx = num.toString();
if(x<=1 && x>=0.50){
    var label = "Website is "+txtx +"% safe to use...";
    document.getElementById("prediction").innerHTML = label;
```

```
document.getElementById("button1").style.display="block";
}
else if (0<=x && x<0.50){
    var label = "Website is "+txtx +"% unsafe to use..."
    document.getElementById("prediction").innerHTML = label;
    document.getElementById("button2").style.display="block";
}
</script>
</body>
```

app.py

#importing required libraries

from flask import Flask, request, render_template import numpy as np import pandas as pd from sklearn import metrics

```
import warnings
import pickle
warnings.filterwarnings('ignore')
from feature import FeatureExtraction
file = open("pickle/model.pkl", "rb")
gbc = pickle.load(file)
file.close()
app = Flask( name )
@app.route("/", methods=["GET", "POST"])
def index():
  if request.method == "POST":
     url = request.form["url"]
     obj = FeatureExtraction(url)
     x = np.array(obj.getFeaturesList()).reshape(1,30)
     y_pred =gbc.predict(x)[0]
     #1 is safe
     #-1 is unsafe
     y_pro_phishing = gbc.predict_proba(x)[0,0]
     y_pro_non_phishing = gbc.predict_proba(x)[0,1]
     # if(y_pred ==1):
     pred = "It is {0:.2f} % safe to go ".format(y_pro_phishing*100)
     return render_template('index.html',xx
=round(y_pro_non_phishing,2),url=url)
  return render_template("index.html", xx =-1)
if __name __ == "__ main __":
  app.run(debug=True)
```

features.py

```
import ipaddress
import re
import urllib.request
from bs4 import BeautifulSoup
import socket
import requests
from googlesearch import search
import whois
from datetime import date, datetime
import time
from dateutil.parser import parse as date_parse
from urllib.parse import urlparse
```

class FeatureExtraction:

```
features = []

def __init__(self,url):
    self.features = []
    self.url = url
    self.domain = ""
    self.whois_response = ""
    self.urlparse = ""
    self.response = ""
    self.soup = ""
```

```
try:
    self.response = requests.get(url)
    self.soup = BeautifulSoup(response.text, 'html.parser')
except:
```

```
pass
```

```
try:
    self.urlparse = urlparse(url)
    self.domain = self.urlparse.netloc
except:
    pass

try:
    self.whois_response = whois.whois(self.domain)
except:
    pass
```

```
self.features.append(self.Usinglp())
self.features.append(self.longUrl())
self.features.append(self.shortUrl())
self.features.append(self.symbol())
self.features.append(self.redirecting())
self.features.append(self.prefixSuffix())
self.features.append(self.SubDomains())
self.features.append(self.Hppts())
self.features.append(self.DomainRegLen())
self.features.append(self.Favicon())
```

```
self.features.append(self.NonStdPort())
self.features.append(self.HTTPSDomainURL())
self.features.append(self.RequestURL())
```

```
self.features.append(self.AnchorURL())
  self.features.append(self.LinksInScriptTags())
  self.features.append(self.ServerFormHandler())
  self.features.append(self.InfoEmail())
  self.features.append(self.AbnormalURL())
  self.features.append(self.WebsiteForwarding())
  self.features.append(self.StatusBarCust())
  self.features.append(self.DisableRightClick())
  self.features.append(self.UsingPopupWindow())
  self.features.append(self.lframeRedirection())
  self.features.append(self.AgeofDomain())
  self.features.append(self.DNSRecording())
  self.features.append(self.WebsiteTraffic())
  self.features.append(self.PageRank())
  self.features.append(self.GoogleIndex())
  self.features.append(self.LinksPointingToPage())
  self.features.append(self.StatsReport())
#1.Usinglp
def Usinglp(self):
  try:
     ipaddress.ip_address(self.url)
     return -1
  except:
     return 1
# 2.longUrl
def longUrl(self):
  if len(self.url) < 54:
```

```
if len(self.url) >= 54 and len(self.url) <= 75:
        return 0
     return -1
  #3.shortUrl
  def shortUrl(self):
     match =
re.search('bit\.ly|goo\.gl|shorte\.st|go2l\.ink|x\.co|ow\.ly|t\.co|tinyurl|tr\.im|i
s\.gd|cli\.gs|'
'yfrog\.com|migre\.me|ff\.im|tiny\.cc|url4\.eu|twit\.ac|su\.pr|twurl\.nl|snipurl
\.com|'
'short\.to|BudURL\.com|ping\.fm|post\.ly|Just\.as|bkite\.com|snipr\.com|fic
\.kr|loopt\.us|'
'doiop\.com|short\.ie|kl\.am|wp\.me|rubyurl\.com|om\.ly|to\.ly|bit\.do|t\.co|l
nkd\.in|'
'db\.tt|gr\.ae|adf\.ly|goo\.g||bitly\.com|cur\.lv|tinyurl\.com|ow\.ly|bit\.ly|ity\.i
m|'
'q\.gs|is\.gd|po\.st|bc\.vc|twitthis\.com|u\.to|j\.mp|buzurl\.com|cutt\.us|u\.b
b|yourls\.org|'
'x\.co|prettylinkpro\.com|scrnch\.me|filoops\.info|vzturl\.com|gr\.net|1url\.c
om|tweez\.me|v\.gd|tr\.im|link\.zip\.net', self.url)
     if match:
        return -1
     return 1
```

return 1

```
#4.Symbol@
def symbol(self):
  if re.findall("@",self.url):
     return -1
   return 1
#5.Redirecting//
def redirecting(self):
  if self.url.rfind('//')>6:
     return -1
   return 1
#6.prefixSuffix
def prefixSuffix(self):
  try:
     match = re.findall('\-', self.domain)
     if match:
        return -1
     return 1
  except:
     return -1
#7.SubDomains
def SubDomains(self):
  dot_count = len(re.findall("\.", self.url))
  if dot_count == 1:
     return 1
  elif dot_count == 2:
     return 0
  return -1
```

```
#8.HTTPS
  def Hppts(self):
     try:
       https = self.urlparse.scheme
       if 'https' in https:
          return 1
       return -1
     except:
       return 1
  #9.DomainRegLen
  def DomainRegLen(self):
     try:
       expiration_date = self.whois_response.expiration_date
       creation_date = self.whois_response.creation_date
       try:
          if(len(expiration_date)):
            expiration_date = expiration_date[0]
       except:
          pass
       try:
          if(len(creation_date)):
            creation_date = creation_date[0]
       except:
          pass
       age = (expiration_date.year-creation_date.year)*12+
(expiration_date.month-creation_date.month)
       if age >=12:
          return 1
```

```
return -1
     except:
        return -1
  # 10. Favicon
  def Favicon(self):
     try:
        for head in self.soup.find_all('head'):
          for head.link in self.soup.find_all('link', href=True):
             dots = [x.start(0) for x in re.finditer('\.', head.link['href'])]
             if self.url in head.link['href'] or len(dots) == 1 or domain in
head.link['href']:
                return 1
        return -1
     except:
        return -1
  # 11. NonStdPort
  def NonStdPort(self):
     try:
        port = self.domain.split(":")
        if len(port)>1:
       return -1
        return 1
     except:
        return -1
  #12. HTTPSDomainURL
  def HTTPSDomainURL(self):
     try:
        if 'https' in self.domain:
```

```
return -1
        return 1
     except:
        return -1
  # 13. RequestURL
  def RequestURL(self):
     try:
        for img in self.soup.find_all('img', src=True):
           dots = [x.start(0) for x in re.finditer('\.', img['src'])]
           if self.url in img['src'] or self.domain in img['src'] or len(dots) ==
1:
             success = success + 1
           i = i + 1
        for audio in self.soup.find_all('audio', src=True):
           dots = [x.start(0) for x in re.finditer('\.', audio['src'])]
          if self.url in audio['src'] or self.domain in audio['src'] or len(dots)
== 1:
             success = success + 1
           i = i + 1
        for embed in self.soup.find_all('embed', src=True):
           dots = [x.start(0) for x in re.finditer('\.', embed['src'])]
           if self.url in embed['src'] or self.domain in embed['src'] or
len(dots) == 1:
             success = success + 1
           i = i + 1
        for iframe in self.soup.find_all('iframe', src=True):
           dots = [x.start(0) for x in re.finditer('\.', iframe['src'])]
```

```
if self.url in iframe['src'] or self.domain in iframe['src'] or
len(dots) == 1:
             success = success + 1
          i = i + 1
        try:
          percentage = success/float(i) * 100
          if percentage < 22.0:
             return 1
          elif((percentage >= 22.0) and (percentage < 61.0)):
             return 0
           else:
             return -1
        except:
          return 0
     except:
        return -1
  # 14. AnchorURL
  def AnchorURL(self):
     try:
        i,unsafe = 0,0
        for a in self.soup.find_all('a', href=True):
          if "#" in a['href'] or "javascript" in a['href'].lower() or "mailto" in
a['href'].lower() or not (url in a['href'] or self.domain in a['href']):
             unsafe = unsafe + 1
          i = i + 1
     try:
          percentage = unsafe / float(i) * 100
```

if percentage < 31.0:

```
return 1
           elif ((percentage >= 31.0) and (percentage < 67.0)):
              return 0
           else:
              return -1
        except:
           return -1
     except:
        return -1
  # 15. LinksInScriptTags
   def LinksInScriptTags(self):
     try:
        i,success = 0,0
     for link in self.soup.find_all('link', href=True):
           dots = [x.start(0) for x in re.finditer('\.', link['href'])]
           if self.url in link['href'] or self.domain in link['href'] or len(dots)
== 1:
              success = success + 1
           i = i + 1
        for script in self.soup.find_all('script', src=True):
           dots = [x.start(0) for x in re.finditer('\.', script['src'])]
           if self.url in script['src'] or self.domain in script['src'] or len(dots)
== 1:
              success = success + 1
           i = i + 1
        try:
```

```
if percentage < 17.0:
             return 1
          elif((percentage >= 17.0) and (percentage < 81.0)):
             return 0
          else:
             return -1
        except:
          return 0
     except:
        return -1
  # 16. ServerFormHandler
  def ServerFormHandler(self):
     try:
        if len(self.soup.find_all('form', action=True))==0:
       return 1
        else:
          for form in self.soup.find_all('form', action=True):
             if form['action'] == "" or form['action'] == "about:blank":
                return -1
             elif self.url not in form['action'] and self.domain not in
form['action']:
                return 0
             else:
                return 1
     except:
        return -1
  # 17. InfoEmail
  def InfoEmail(self):
```

percentage = success / float(i) * 100

```
try:
     if re.findall(r"[mail\(\)|mailto:?]", self.soap):
        return -1
     else:
        return 1
  except:
     return -1
# 18. AbnormalURL
def AbnormalURL(self):
  try:
     if self.response.text == self.whois_response:
        return 1
     else:
        return -1
  except:
     return -1
#19. WebsiteForwarding
def WebsiteForwarding(self):
  try:
     if len(self.response.history) <= 1:
        return 1
     elif len(self.response.history) <= 4:
        return 0
     else:
        return -1
  except:
      return -1
```

20. StatusBarCust

```
def StatusBarCust(self):
     try:
       if re.findall("<script>.+onmouseover.+</script>",
self.response.text):
          return 1
        else:
          return -1
     except:
        return -1
  #21. DisableRightClick
  def DisableRightClick(self):
     try:
       if re.findall(r"event.button ?== ?2", self.response.text):
          return 1
        else:
      return -1
     except:
        return -1
  # 22. UsingPopupWindow
  def UsingPopupWindow(self):
     try:
       if re.findall(r"alert\(", self.response.text):
          return 1
        else:
          return -1
     except:
        return -1
```

23. IframeRedirection

```
def IframeRedirection(self):
     try:
       if re.findall(r"[<iframe>|<frameBorder>]", self.response.text):
          return 1
       else:
          return -1
     except:
        return -1
  #24. AgeofDomain
  def AgeofDomain(self):
     try:
       creation_date = self.whois_response.creation_date
    try:
          if(len(creation_date)):
            creation_date = creation_date[0]
       except:
          pass
       today = date.today()
       age = (today.year-creation_date.year)*12+(today.month-
creation_date.month)
       if age >=6:
          return 1
       return -1
     except:
       return -1
  #25. DNSRecording
  def DNSRecording(self):
     try:
```

```
creation_date = self.whois_response.creation_date
       try:
          if(len(creation_date)):
            creation_date = creation_date[0]
       except:
          pass
       today = date.today()
       age = (today.year-creation_date.year)*12+(today.month-
creation_date.month)
       if age >=6:
          return 1
       return -1
     except:
     return -1
  # 26. WebsiteTraffic
  def WebsiteTraffic(self):
     try:
       rank =
BeautifulSoup(urllib.request.urlopen("http://data.alexa.com/data?cli=10&
dat=s&url=" + url).read(), "xml").find("REACH")['RANK']
       if (int(rank) < 100000):
          return 1
       return 0
     except:
       return -1
  #27. PageRank
  def PageRank(self):
     try:
```

```
prank_checker_response =
requests.post("https://www.checkpagerank.net/index.php", {"name":
self.domain))
       global_rank = int(re.findall(r"Global Rank: ([0-9]+)",
rank_checker_response.text)[0])
       if global_rank > 0 and global_rank < 100000:
          return 1
        return -1
     except:
       return -1
 # 28. GoogleIndex
  def GoogleIndex(self):
                site = search(self.url, 5)
     try:
       if site:
          return 1
       else:
                       return -1
     except:
       return 1
  #29. LinksPointingToPage
  def LinksPointingToPage(self):
     try:
       number_of_links = len(re.findall(r"<a href=", self.response.text))</pre>
       if number_of_links == 0:
          return 1
       elif number_of_links <= 2:
          return 0
                       return -1
       else:
```

#30. StatsReport def StatsReport(self): url_match = re.search(try: 'at\.ua|usa\.cc|baltazarpresentes\.com\.br|pe\.hu|esy\.es|hol\.es|sweddy\. com|myjino\.ru|96\.lt|ow\.ly', url) ip_address = socket.gethostbyname(self.domain) ip match = e.search('146\.112\.61\.108|213\.174\.157\.151|121\.50\.168\.88|192\.18 5\.217\.116|78\.46\.211\.158|181\.174\.165\.13|46\.242\.145\.103|121\.50 \.168\.40|83\.125\.22\.219|46\.242\.145\.98|' '107\.151\.148\.44|107\.151\.148\.107|64\.70\.19\.203|199\.184\.144\.27| 107\.151\.148\.108|107\.151\.148\.109|119\.28\.52\.61|54\.83\.43\.69|52\. 69\.166\.231|216\.58\.192\.225|' '118\.184\.25\.86|67\.208\.74\.71|23\.253\.126\.58|104\.239\.157\.210|17 5\.126\.123\.219|141\.8\.224\.221|10\.10\.10\.10|43\.229\.108\.32|103\.23 2\.215\.140|69\.172\.201\.153|' '216\.218\.185\.162|54\.225\.104\.146|103\.243\.24\.98|199\.59\.243\.120 |31\.170\.160\.61|213\.19\.128\.77|62\.113\.226\.131|208\.100\.26\.234|1 95\.16\.127\.102|195\.16\.127\.157|' '34\.196\.13\.28|103\.224\.212\.222|172\.217\.4\.225|54\.72\.9\.51|192\.6 4\.147\.141|198\.200\.56\.183|23\.253\.164\.103|52\.48\.191\.26|52\.214\. 197\.72|87\.98\.255\.18|209\.99\.17\.27|' '216\.38\.62\.18|104\.130\.124\.96|47\.89\.58\.141|78\.46\.211\.158|54\.8 6\.225\.156|54\.82\.156\.19|37\.157\.192\.102|204\.11\.56\.48|110\.34\.23 1\.42', ip_address) if url_match: return -1 elif ip_match: return -1

return -1

except:

return 1

except: return 1

def getFeaturesList(self):
 return self.features

8. TESTING

8.1 Test Cases

The experimental system is structured to empirically test and verify the efficacy of the proposed methods for phishing website detection. For training and evaluating the proposed techniques, three phishing datasets from the UCI repositories are used and the K-fold (where k = 10) cross-validation (CV) approach is used for the creation and evaluation of the phishing models.

The 10-fold CV option is based on its ability to create phishing models with the low impact of the issue of class imbalance [49, 50, 51, 52, 53]. Moreover, the K-fold CV approach ensures that each instance can be used iteratively for both training and testing [54, 55, 56].

On phishing datasets, based on 10-fold CV, the proposed methods and the chosen baseline classifiers (NB, SMO, SVM, and Decision Table (Dec Table)) are then implemented.

The phishing detection efficiency of the developed phishing models is then tested and contrasted with other experimented methods of phishing detection. All experiments were performed using the WEKA machine learning tool in the same environment [57].

8.2 User Acceptance Testing

Phishing attacks around the world cost billions of dollars in loss every year (Mc Combie et.al, 2009). Phishing has a huge negative impact on organizations' client relationships revenues, marketing pains and general corporate appearance (Dhanalakshmi et.al, 2011). Statistics report that 35.9% of financial sector is the target of Phishing frauds (APWG, 2010).

However these approaches are cannot detect and identify fresh phishes because of lists, where maintenance and human resources required and the scalability and run time are not suitable. This is the reason the list based approaches combine with other approaches [2, 3, 10,

[12][13][14][15][16][17][18][19][20].

The Heuristics based approaches are predicted through one or more websites features like URL, source code and visual features.

9. RESULTS 9.1 Performance Matrix

Coding For Metrics:

```
import pandas as pd
import numpy as np
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import confusion_matrix,accuracy_score
```

```
#Import Dataset
ds= pd.read_csv("dataset_website.csv")
ds.head()
```

```
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression()
lr.fit(x_train,y_train)
```

```
y_pred1=lr.predict(x_test)|
from sklearn.metrics import accuracy_score
log_reg=accuracy_score(y_test,y_pred1)
log_reg
```

0.9167797376752601

	Index	having_IPhaving_IP_Address	URLURL_Length	Shortining_Service	having_At_Symbol	double_slash_redirecting	Prefix_Suffix	having_Sub_Domain
0	1	-1	1	1	-1	-1	-1	-1
1	2	1	1	1	1	1	-1	c
2	3	1	0	1	1	1	-1	-1
3	4	1	0	1	1	1	-1	-1
4	5	1	0	-1	1	1	-1	1

5 rows × 32 columns

```
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression()
lr.fit(x_train,y_train)
```

```
import pickle
pickle.dump(lr,open('Phishing_Website.pkl','wb'))
```

```
import numpy as np
from flask import Flask, request, jsonify, render_template
import pickle
#importing the inputScript file used to analyze the URL
import inputScript
```

```
#load model
g app = Flask(__name__)
model = pickle.load(open('Phishing_Website.pkl', 'rb'))
11
```

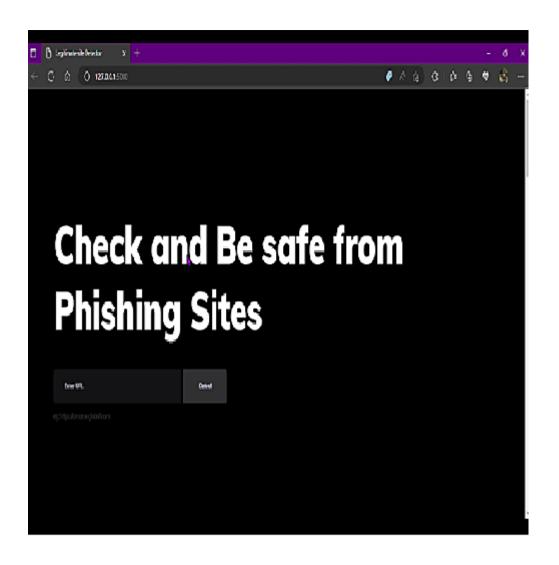
```
#Redirects to the page to give the user iput URL. @app.route('/predict')
    * def predict():
          return render template('final.html')
      #Fetches the URL given by the URL and passes to inputScript @app.route('/y_predict',methods=['POST'])
     def y_predict():
           For rendering results on HTML GUI
          url = request.form['URL']
checkprediction = inputScript.main(url)
prediction = model.predict(checkprediction)
           print(prediction)
          output=prediction[0]
           if(output==1):
               pred="Your are safe!! This is a Legitimate Website."
               pred="You are on the wrong site. Be cautious!"
           return render_template('final.html', prediction_text='{}'.format(pred),url=url)
      #Takes the input parameters fetched from the URL by inputScript and returns the predictions
      @app.route('/predict_api',methods=['POST'])
    ▼ def predict_api():
           For direct API calls trought request
          data = request.get_json(force=True)
          prediction = model.y_predict([np.array(list(data.values()))])
44
          output = prediction[0]
           return jsonify(output)
```

```
51

52 if __name__ == '__main__':

53 app.run(host='0.0.0.0', debug=True)

54
```



10 .Advantages And Disadvantages

Advantages:

- Could Easily find the phishing websites before clicking on them
- 2. Can Identify Which Sites Have The Most Target
- 3. Could save your data before it is too late
- 4. DETECTION is Better Than Cure.

Disadvantages:

- 5. Hackers Find New Way To Attack,
- 6. May not be able to detect all the websites.
- 7. Could possibly bypass the detection.

11.Conclusion

The importance to safeguard online users from becoming victims of online fraud, divulging confidential information to an attacker among other effective uses of phishing as an attacker's tool, phishing detection tools play a vital role in ensuring a secure online experience for users.

Unfortunately, many of the existing phishing-detection tools, especially those that depend on an existing blacklist, suffer limitations such as low detection accuracy and high false alarm that is often caused by either a delay in blacklist update as a result of human <u>verification process</u> involved in <u>classification</u> or perhaps, it can be attributed to human error in classification which may lead to improper classification of the classes.

These critical issues have drawn many researchers to work on various approaches to improve detection accuracy of phishing attacks and to minimize false alarm rate. The inconsistent nature of attacks behaviors and continuously changing URL phish patterns require timely updating of the reference model.

Therefore, it requires an effective technique to regulate retraining as to enable <u>machine learning algorithm</u> to actively adapt to the changes in phish patterns.

12. Future Scope

Despite there are several ways to carry out these attacks, unfortunately the current phishing detection techniques cover some attack vectors like email and fake websites. Therefore, building a specific limited scope detection system will not provide complete protection from the wide phishing attack vectors.

13. Source Code

```
HTML
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-</pre>
scale=1.0">
  <meta name="description" content="This website is develop for
identify the safety of url.">
  <meta name="keywords" content="phishing url,phishing,cyber</pre>
security, machine learning, classifier, python">
  <!-- BootStrap -->
  k rel="stylesheet"
href="https://stackpath.bootstrapcdn.com/bootstrap/4.5.0/css/bootstrap.
min.css"
     integrity="sha384-
9alt2nRpC12Uk9gS9baDl411NQApFmC26EwAOH8WgZl5MYYxFfc+Nc
Pb1dKGj7Sk" crossorigin="anonymous">
```

<link href="static/styles.css" rel="stylesheet">

<title>URL detection</title>

</head>

```
<body>
<div class=" container">
  <div class="row">
     <div class="form col-md" id="form1">
       <h2>PHISHING URL DETECTION</h2>
       <hr>
       <form action="/" method ="post">
          <input type="text" class="form__input" name ='url' id="url"</pre>
placeholder="Enter URL" required="" />
          <label for="url" class="form label">URL</label>
          <button class="button" role="button" >Check here
       </form>
  </div>
  <div class="col-md" id="form2">
     <br>
     <h6 class = "right "><a href= {{ url }} target="_blank">{{ url
}}</a></h6>
     <br>
     <h3 id="prediction"></h3>
     <button class="button2" id="button2" role="button"</pre>
onclick="window.open('{{url}}')" target="_blank" >Still want to
Continue</button>
     <button class="button1" id="button1" role="button"</pre>
onclick="window.open('{{url}}')" target="_blank">Continue</button>
```

```
</div>
</div>
<br>
</div>
  <!-- JavaScript -->
  <script src="https://code.jquery.com/jquery-3.5.1.slim.min.js"</pre>
    integrity="sha384-
DfXdz2htPH0lsSSs5nCTpuj/zy4C+OGpamoFVy38MVBnE+lbbVYUew+
OrCXaRkfj"
    crossorigin="anonymous"></script>
  <script
src="https://cdn.jsdelivr.net/npm/popper.js@1.16.0/dist/umd/popper.min.j
s"
    integrity="sha384-
Q6E9RHvblyZFJoft+2mJbHaEWldlvl9IOYy5n3zV9zzTtml3UksdQRVvox
MfooAo"
    crossorigin="anonymous"></script>
  <script
src="https://stackpath.bootstrapcdn.com/bootstrap/4.5.0/js/bootstrap.min
.js"
    integrity="sha384-
OgVRvuATP1z7JjHLkuOU7Xw704+h835Lr+6QL9UvYjZE3Ipu6Tp75j7B
h/kR0JKI"
    crossorigin="anonymous"></script>
  <script>
       let x = '{\{xx\}}';
```

```
let num = x*100;
       if (0 \le x \&\& x \le 0.50)
          num = 100-num;
       }
       let txtx = num.toString();
       if(x \le 1 \&\& x \ge 0.50)
          var label = "Website is "+txtx +"% safe to use...";
          document.getElementById("prediction").innerHTML = label;
          document.getElementById("button1").style.display="block";
       }
       else if (0 \le x \&\& x < 0.50){
          var label = "Website is "+txtx +"% unsafe to use..."
          document.getElementById("prediction").innerHTML = label;
          document.getElementById("button2").style.display="block";
       }
  </script>
</body>
</html>
```

TEAM GITHUB LINK: https://github.com/IBM-EPBL/IBM-Project-5743-1658813865

Demo Video: