

# A MACHINE LEARNING- BASED FRAMEWORK FOR PHISHING DETECTION AND PREVENTION

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# INTRODUCTION

# WHAT IS MACHINE LEARNING?

## Definition:

- A branch of artificial intelligence focused on enabling computers to learn from data.

## Purpose:

- Automate decision-making and predictions based on patterns in data.

## Types:

- Supervised, unsupervised, and reinforcement learning.

# WHAT IS PHISHING?

## Definition:

- A type of cyber-attack where attackers attempt to deceive users into providing sensitive information.

## Methods:

- Often involves fraudulent emails, messages, or websites.

## Impact:

- Can lead to identity theft, financial loss, or unauthorized access to data.

## DANGERS OF PHISHING

22%

FBI'S 2021 INTERNET  
CRIME REPORT  
SHOWS PHISHING  
ATTACKS ACCOUNTED  
FOR 22% OF ALL  
DATA BREACHES

\$44.2 M

\$44.2 MILLION WAS  
STOLEN BY CYBER  
CRIMINALS  
THROUGH PHISHING  
ATTACKS IN 2021.  
(AAG IT)

97%

INTEL STUDY  
SHOWS 97% OF  
PEOPLE FAIL AT  
IDENTIFYING  
PHISHING EMAILS  
FROM GENUINE  
EMAILS

## OUR SOLUTION

The overarching problem addressed by this project is the persistent vulnerability of individuals and organizations to phishing threats.

We set out to design a sophisticated AI model-based tool aimed at enhancing cybersecurity and protecting users through the detection of link-related phishing attacks in web content



# METHODOLOGY



## Research the problem

Understand the intricacies of phishing attacks, their indicators, and machine learning models



## Find the data

Gather comprehensive datasets containing benign and phishing links, ensuring that the data is diverse and up-to-date.



## Build the model

Utilize machine learning algorithms to construct a classifier that can accurately distinguish between safe and phishing links.



## Develop the tool

Integrate the model into a user-friendly application, providing real-time phishing detection to assist users in staying secure online.



# DATASET



# DATASET ACQUISITION AND PREPARATION

- FINAL DATASETS BUILT UPON:
  - GREGA VRBANČIČ'S [HTTPS://DOI.ORG/10.1016/J.DIB.2020.106438](https://doi.org/10.1016/J.DIB.2020.106438)
  - ABDELHAKIM HANNOUSSE AND SALIMA YAHIOUCHE'S [HTTPS://DOI.ORG/10.17632/C2GW7FY2J4.3](https://doi.org/10.17632/C2GW7FY2J4.3)
- DATASETS ARE CLEANED, TRANSFORMED, ENRICHED, MERGED, AND TRIMMED USING PANDAS.
- 65.36% NON-PHISHING LINKS, 34.64% PHISHING LINKS
- SHAPE: (40,000, 116)
  - 40,000 INSTANCES
  - 116 FEATURES

# URL FEATURIZATION



- Length of the URL and its sub-components
- Quantity of special characters and their ratios in different parts of the URL
- IP address as domain
- Redirections
- Suspicious keywords within sub-components
- HTTPS as the protocol or a part of another section of the URL

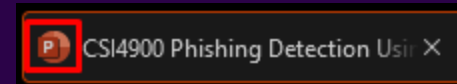
# DOMAIN FEATURIZATION

- Indexed by Google
- Time response
- TLS/SSL Certification
- Page Rank (Tranco)
- DNS Record
- Domain Age
- Quantity of Resolved IPs
- Quantity of servers
  - MX servers
  - Name servers

# ROOM FOR IMPROVEMENT

## HTML/JAVASCRIPT ELEMENTS FROM URL PAGES

- Existence of Favicon
- Javascript Obfuscation
- Extraction of HTML tags from the page
  - <script>, <meta>, <link>, <a>



(A)

```
function setText(data) {  
  document.getElementById("myDiv").innerHTML = data;  
}
```

(B)

```
function ghds3x(n) {  
  h = "\x69\u0065\u0065r\x48T\u004DL";  
  a="s c v o v d h e , n i";x=a.split(" ");b="gztXleWentBsyf";  
  r=b.replace("z",x[7]).replace("x","E").replace("s","").replace("f","I")  
  ["repl" + "ace"]("W","m")+ "d";  
  c="my"+String.fromCharCode(68)+x[10]+"v";  
  s=x[5]+x[3]+x[1]+"um"+x[7]+x[9]+"t";d=this[s][r](c);if(+!![])  
  { d[h]=n; } else { d[h]=c; } }
```

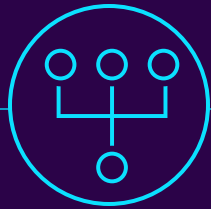
# MODEL

# MODEL DEVELOPMENT



## Research

Researched many potential methods to handle the phishing classification problem



## Model Selection

Numerous models pitted against each other to find best performance



## Featurization & Analysis

Features chosen and analyzed for maximum impact and coverage



## Final Implementation

Implement & combine everything together to construct model ensuring reliable and efficient detection

# RESEARCH

Machine learning models and their training algorithms			
Supervised learning	Unsupervised learning	Semi-supervised learning	Reinforcement learning
<p>Data scientists provide input, output and feedback to build model (as the definition).</p> <p>EXAMPLE ALGORITHMS:</p> <p><b>Linear regressions</b></p> <ul style="list-style-type: none"><li>• Sales forecasting.</li><li>• Risk assessment.</li></ul> <p><b>Support vector machines</b></p> <ul style="list-style-type: none"><li>• Image classification.</li><li>• Financial performance comparison.</li></ul> <p><b>Decision trees</b></p> <ul style="list-style-type: none"><li>• Predictive analytics.</li><li>• Pricing.</li></ul>	<p>Use deep learning to arrive at conclusions and patterns through unlabeled training data.</p> <p>EXAMPLE ALGORITHMS:</p> <p><b>Apriori</b></p> <ul style="list-style-type: none"><li>• Sales functions.</li><li>• Word associations.</li><li>• Searcher.</li></ul> <p><b>K-means clustering</b></p> <ul style="list-style-type: none"><li>• Performance monitoring.</li><li>• Searcher intent.</li></ul> <p><b>Artificial neural networks</b></p> <ul style="list-style-type: none"><li>• Generate new, synthetic data.</li><li>• Data mining and pattern recognition.</li></ul>	<p>Builds a model through a mix of labeled and unlabeled data, a set of categories, suggestions and example labels.</p> <p>EXAMPLE ALGORITHMS:</p> <p><b>Generative adversarial networks</b></p> <ul style="list-style-type: none"><li>• Audio and video manipulation.</li><li>• Data creation.</li></ul> <p><b>Self-trained Naïve Bayes classifier</b></p> <ul style="list-style-type: none"><li>• Natural language processing.</li></ul>	<p>Self-interpreting but based on a system of rewards and punishments learned through trial and error, seeking maximum reward.</p> <p>EXAMPLE ALGORITHMS:</p> <p><b>Q-learning</b></p> <ul style="list-style-type: none"><li>• Policy creation.</li><li>• Consumption reduction.</li></ul> <p><b>Model-based value estimation</b></p> <ul style="list-style-type: none"><li>• Linear tasks.</li><li>• Estimating parameters.</li></ul>

Fig 1. The four main types of machine learning and their most common algorithms.

## Languages and Packages:

- Many languages and Packages to choose from
- Python using scikit-learn as main foundation found to be most suitable

## Feature Types:

- Various possible types such as text-based, URL-based, HTML-Based, signature-based, etc.
- Chose URL-based with some signature-based mixed in

## Classification Methods:

- Numerous methods such as supervised, unsupervised, etc.
- A Supervised method chosen to compliment URL-based features

## Phishing Indicators

- Many types of attacks
- What signals a possible URL-based attack?
- Common metrics like checking google page rank or URL shortening services



# MODEL SELECTION

	PRECISION	RECALL	F1-SCORE	TEST SCORE	TRAIN SCORE
Random Forest	0   0.97 1   0.94	0   0.97 1   0.95	0   0.97 1   0.94	0.961	0.999
Light GBM	0   0.97 1   0.94	0   0.97 1   0.95	0   0.97 1   0.94	0.960	0.977
Logistic Regression	0   0.95 1   0.87	0   0.93 1   0.91	0   0.94 1   0.89	0.922	0.924
Decision Tree	0   0.95 1   0.92	0   0.96 1   0.91	0   0.96 1   0.92	0.944	1
Naïve Bayes	0   0.92 1   0.80	0   0.89 1   0.86	0   0.91 1   0.83	0.879	0.879
Support Vector Machine	0   0.70 1   0.81	0   0.98 1   0.20	0   0.81 1   0.31	0.707	0.707
Multilayer Perceptron	0   0.83 1   0.95	0   0.98 1   0.63	0   0.90 1   0.75	0.859	0.862
BEST	Random Forest	Random Forest	Random Forest	Random Forest	Decision Tree

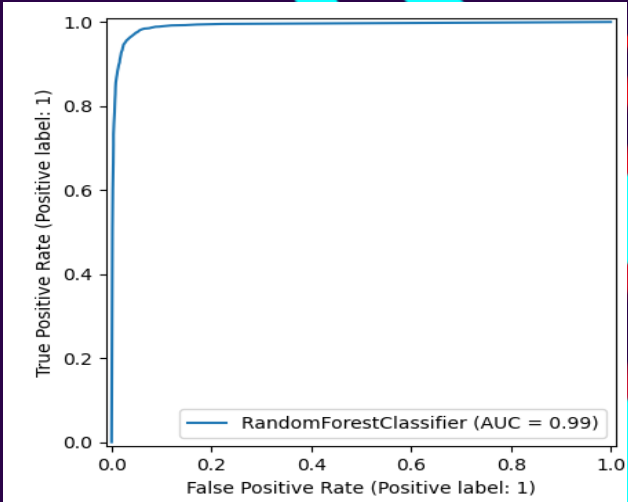


Fig 2. Random Forest AUC/ROC Display

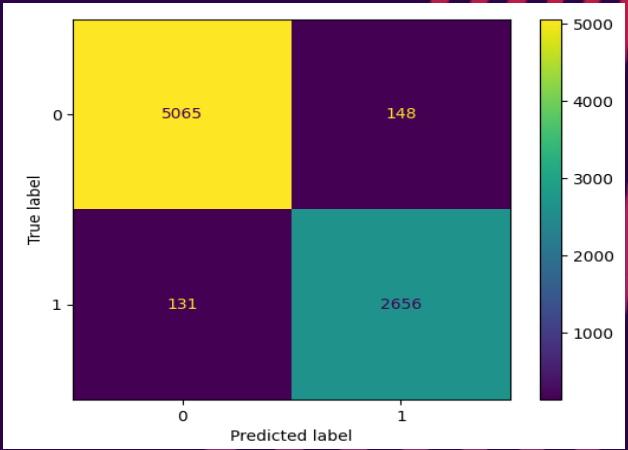


Fig 3. Random Forest Confusion Matrix

# FEATURE ANALYSIS

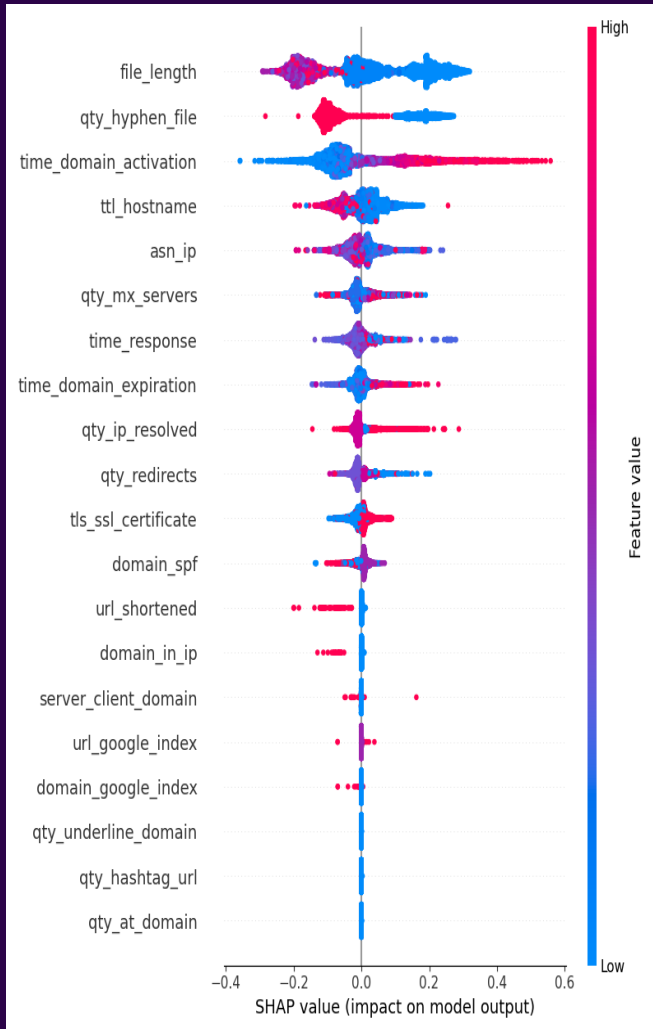


Fig 4. SHAP Summary Graph

## Variance Inflation Factor (VIF):

- Assess multicollinearity among features to ensure that each feature provides unique information to the model.

## SHAP:

- values to understand each feature's contribution to the model's predictions and their impact on classification outcomes.

## Permutation Importance:

- Measure the importance of each feature by observing the change in model performance when the feature's values are randomly permuted.

## LIME

- provide interpretable insights into the model's predictions for specific instances by approximating the model with a simpler, interpretable model.

directory_length	0.180 +/- 0.003
time_domain_activation	0.062 +/- 0.002
length_url	0.016 +/- 0.001
qty_dot_domain	0.010 +/- 0.001
ttn_hostname	0.007 +/- 0.001
qty_nameservers	0.006 +/- 0.001
asn_ip	0.006 +/- 0.001
time_response	0.005 +/- 0.001
qty_mx_servers	0.003 +/- 0.000
qty_ip_resolved	0.003 +/- 0.000
time_domain_expiration	0.003 +/- 0.001
domain_spf	0.002 +/- 0.000

Fig 5. Permutation Importance

Fig 7. LIME Feature Insights

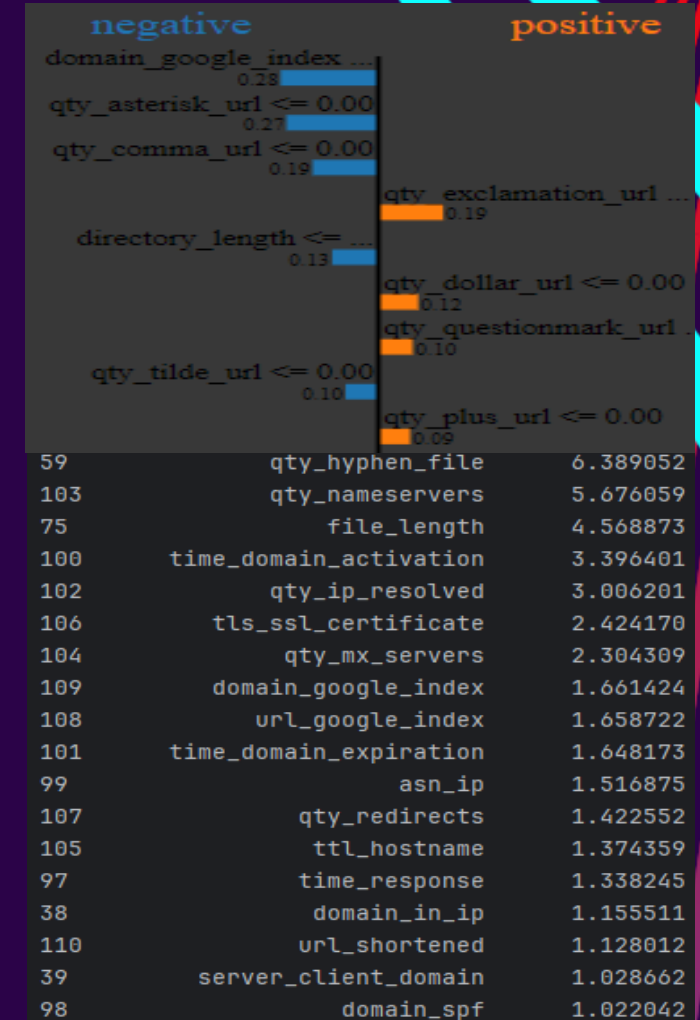


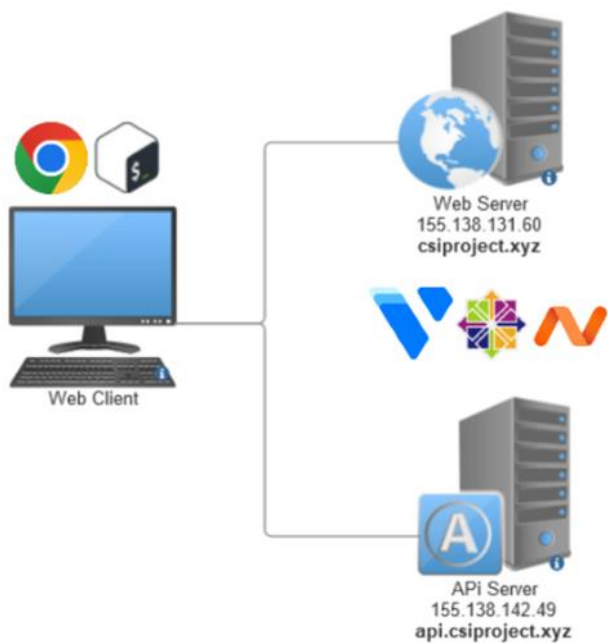
Fig 6. VIF Inflation Factor

## FINAL IMPLEMENTATION

- IMPLEMENTED USING PYTHON, INCLUDING LIBRARIES AND PACKAGES SUCH AS SCIKIT-LEARN, PANDAS, ETC.
- 900+ LINES OF CODE WITH FEATURE IMPLEMENTATION
- TRAINED ON FINAL PROCESSED QUALITY DATASET
- 115 FEATURES, 1 TARGET
- HYPERPARAMETER TUNING FOR BEST RESULTS
- 10-FOLD CROSS VALIDATED, 0.91 MATTHEWS COEFFICIENT
- 96% ACCURACY, F1-SCORE, PRECISION, RECALL ON AVERAGE
- STRONG, RELIABLE AND EFFICIENT MODEL FOR DETECTION

# WEB INFRASTRUCTURE

# OUR SERVERS



## VULTR Provides:

- Hardware Resources for computation  
Toronto DC - Regular Cloud Compute 1 vCPU,  
1024 MB RAM, 25 GB SSD, 1.00 TB Transfer  
(5\$/mo)
- DNS (supporting SLD)
- Firewall
- Reserved IPs
- Snapshots and Optional Upgrades

## Namecheap provides:

- TLD – csiprject.xyz

## Shared configurations

- Centos Stream 9
- Certbot certificates
- Firewall ports opened

Additional 15% Total Sales Tax applicable to this order [Update Settings](#)



Dashboard

Domains → [Details](#)

Expiring / Expired

Domain List

Hosting List

Private Email










SSL Certificates

Apps

Profile

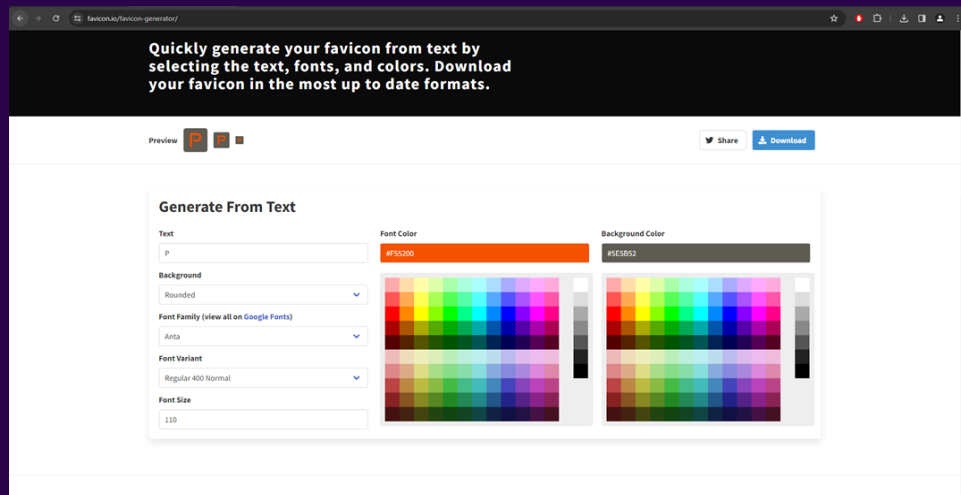
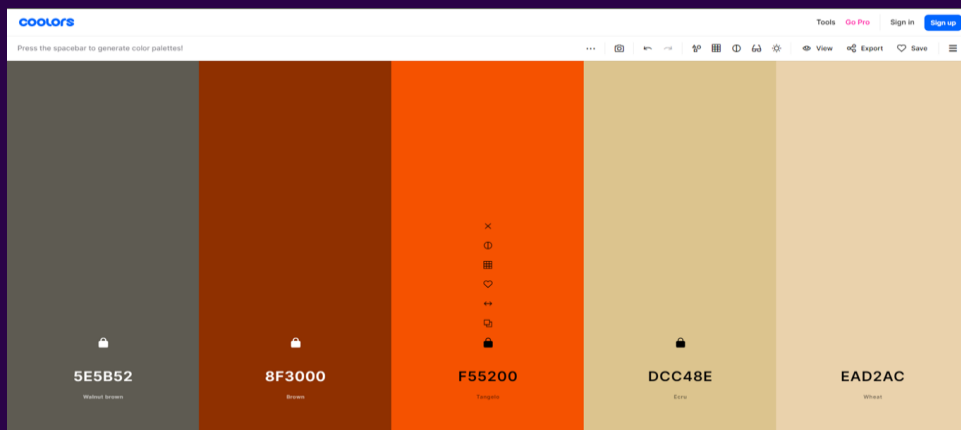


**csiproject.xyz**

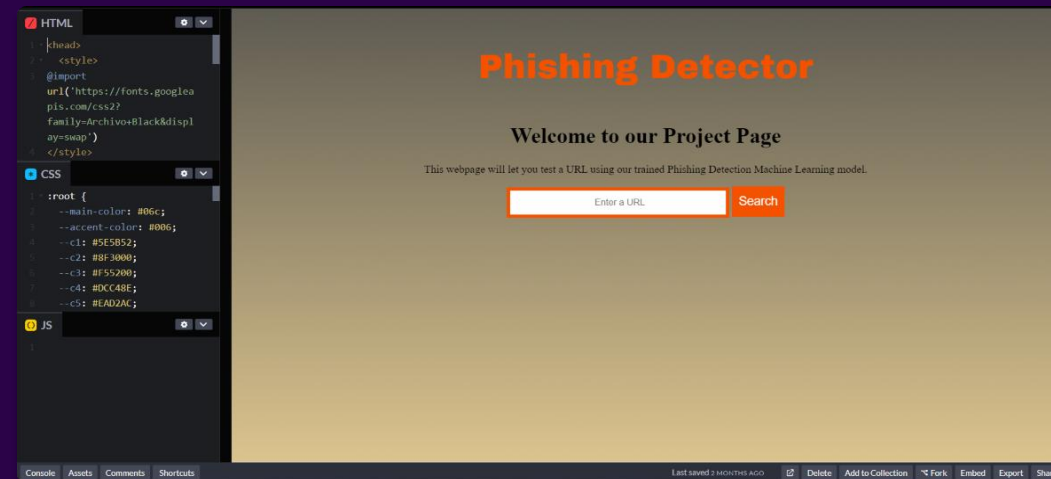
	Domain	Products	Sharing & Transfer	Advanced DNS	
STATUS & VALIDITY	 <input checked="" type="checkbox"/> ACTIVE	Feb 25, 2024 - Feb 25, 2025	<input type="checkbox"/> AUTO-RENEW	ADD YEARS	
 Withheld for Privacy	 <input type="checkbox"/> PROTECTION	Feb 25, 2024 - Feb 25, 2025	<input type="checkbox"/> AUTO-RENEW	ADD YEARS	 SHOW DETAILS
 PremiumDNS	 Enable PremiumDNS protection in order to switch your domain to our PremiumDNS platform. With our PremiumDNS platform, you get 100% DNS uptime and DDoS protection at the DNS level.			BUY NOW	
NAMESERVERS	 Custom DNS <div>           ns1.vultr.com           ns2.vultr.com            ADD NAMESERVER         </div>				
REDIRECT DOMAIN	 You can create redirects via your DNS provider or your Namecheap account. To perform this function from your account, you must first change your nameservers to Namecheap default. <a href="#">Learn How →</a>				



# DESIGN



- Start with basic HTML, CSS, JS
- Keep it simple and consistent
- Introducing more elements adds complexity

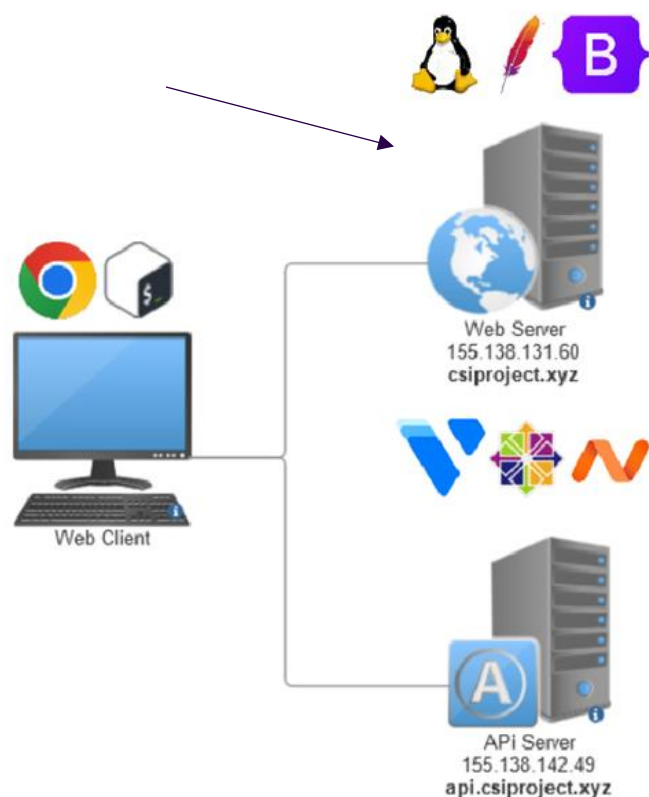


# DESIGN

Figma for prototyping



# WEB SERVER DETAILS



- Using [Apache](#) as the Web Server app
- Sending our webpage files which now incorporate the [Bootstrap](#) framework
- Apache configured to use [HTTPS](#) (port 443) and 303 Redirect (http -> https)
- Self-signed certificate -> [Certbot](#)

These are important for modern websites as web browsers will often complain

```
[root@csiprojwebserver ~]# systemctl status httpd
● httpd.service - The Apache HTTP Server
  Loaded: loaded (/usr/lib/systemd/system/httpd.service; enabled; preset: disabled)
  Active: active (running) since Thu 2024-03-21 00:47:54 UTC; 3 weeks 5 days ago
```

# PHISHING DETECTION AND PREVENTION



- ```
[root@csiprojapi pythonproj]# python api.py
* Serving Flask app 'api'
* Debug mode: on
INFO:werkzeug:WARNING: This is a development server.
* Running on all addresses (0.0.0.0)
* Running on https://127.0.0.1:443
* Running on https://155.138.142.49:443
INFO:werkzeug:Press CTRL+C to quit
```



# RESULTS

Document

https://csiproject.xyz

Phishing Detection Project

Home Overview Datasets ML Model Web Infrastructure Login

## Phishing Detector

Welcome to our CSI4900 Project Page

This webpage will let you test a URL using our trained Machine Learning model for Phishing Detection

Submit

### Prediction Results

| ID | Time                  | URL                                 | Result |
|----|-----------------------|-------------------------------------|--------|
| 2  | 4/15/2024, 1:54:35 AM | https://en.wikipedia.org/wiki/Linux | 0      |
| 1  | 4/15/2024, 1:54:29 AM | https://csiproject.xyz/             | 0      |

Page Info — https://csiproject.xyz/

General Media Permissions Security

Website Identity

Website: csiproject.xyz

Owner: This website does not supply ownership information.

Verified by: Let's Encrypt View Certificate

Privacy & History

Have I visited this website prior to today? Yes, 16 times

Is this website storing information on my computer? No Clear Cookies and Site Data

Have I saved any passwords for this website? No View Saved Passwords

Technical Details

Connection Encrypted (TLS\_AES\_256\_GCM\_SHA384, 256 bit keys, TLS 1.3)

The page you are viewing was encrypted before being transmitted over the Internet.

Encryption makes it difficult for unauthorized people to view information travelling between computers. It is therefore unlikely that anyone read this page as it travelled across the network.

Help

Inspector Console Debugger Network Style Editor Performance Memory Storage Accessibility Application

Filter URLs

| Status | Method | Domain             | File                    | Initiator                       | Type | Transferred | Size     | 0 ms   |
|--------|--------|--------------------|-------------------------|---------------------------------|------|-------------|----------|--------|
| 304    | GET    | csiproject.xyz     | /                       | document                        | html | cached      | 19.54 kB | 205 ms |
| 200    | GET    | csiproject.xyz     | main1.js                | script                          | js   | cached      | 0 B      | 0 ms   |
| 200    | GET    | cdn.jsdelivr.net   | bootstrap.bundle.min.js | script                          | js   | cached      | 0 B      | 0 ms   |
| 200    | GET    | csiproject.xyz     | logo.png                | FaviconLoader.sys.mjs:175 (img) | png  | cached      | 3.51 kB  | 0 ms   |
| 200    | GET    | api.csiproject.xyz | predicetest             | main1.js:61 (xhr)               | html | 240 B       | 1 B      | 206 ms |
| 200    | GET    | api.csiproject.xyz | predicetest             | main1.js:61 (xhr)               | html | 240 B       | 1 B      | 227 ms |

6 requests 23.05 kB / 480 B transferred Finish: 6.81 s DOMContentLoaded: 162 ms load: 163 ms

Filter Output

Errors Warnings Logs Info Debug CSS XHR Requests

Request for font "Noto Sans" blocked at visibility level 2 (requires 3)

Request for font "Liberation Sans" blocked at visibility level 2 (requires 3)

Button was clicked!

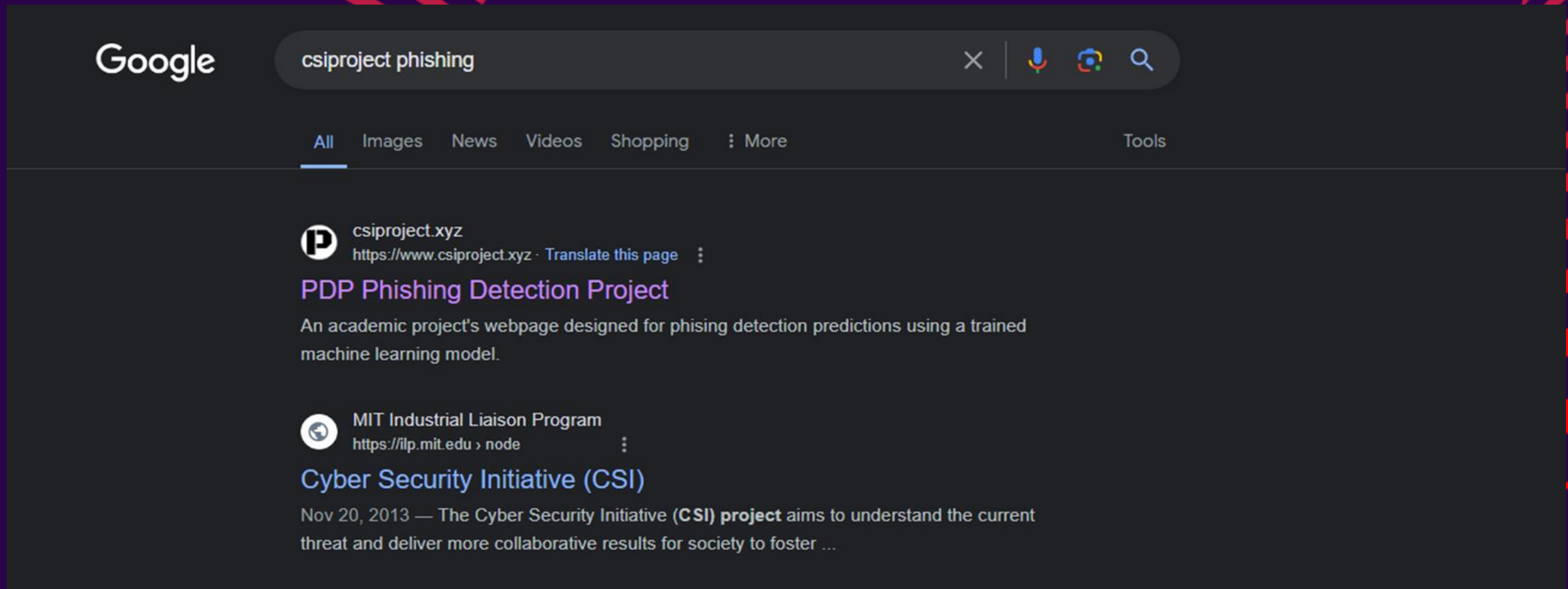
false

Synchronous XMLHttpRequest on the main thread is deprecated because of its detrimental effects to the end user's experience. For more help https://xhr.spec.whatwg.org/#sync-warning

200

Button was clicked!

# RESULTS





## CONCLUSION

Through the design and implementation of a machine learning model and tool for phishing detection, we aimed to help combat the ever-evolving cyber security threats that impact individuals and businesses alike.



# THANK YOU

Are there any questions?