



*King Fahd University of Petroleum and Mineral*  
*The Department of Information and Computer Science*

# Experimentation and Analysis of Present Solutions: Phishpedia

Experimental Report for Phishpedia

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# Agenda

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- ▶ Environment Setup
- ▶ Phishapedia Overview
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- ▶ Challenges and Limitations
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# Introduction



## **What is Phishpedia?**

- A phishing detection tool using deep learning to analyze logos and webpage features.
- Protects users by identifying phishing websites based on visual elements.

## **Objective of Experimentation:**

Test its performance and analyze effectiveness.



# Environment Setup

## **Tools:**

- Kali Linux on VirtualBox.
- Dependencies: Python3, pip3, net-tools, and Anaconda.

## **Installation Steps:**

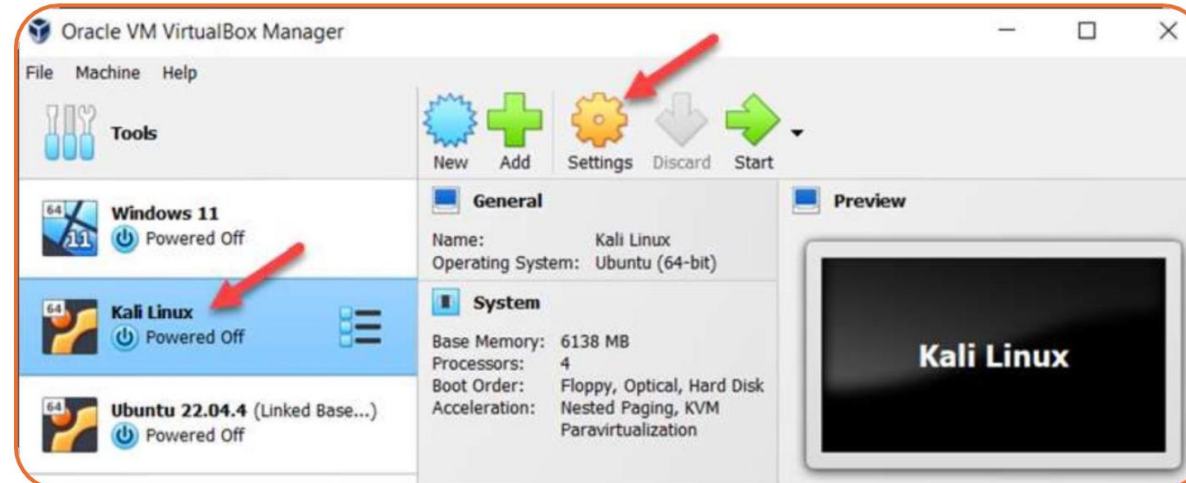
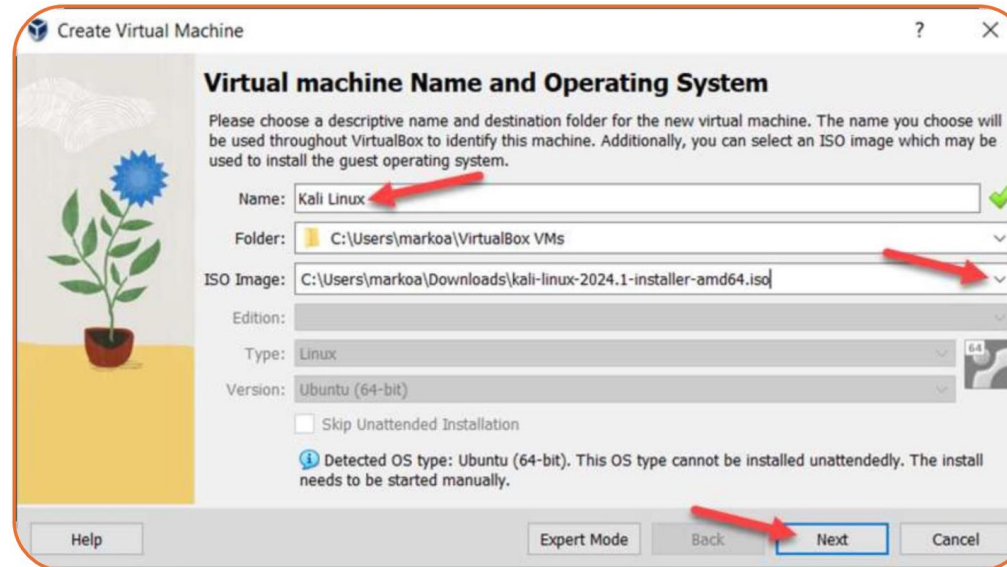
1. Download and install Kali Linux ISO.
2. Configure VirtualBox settings.
3. Install necessary dependencies.

## **Why This Setup?**

To create a controlled environment for experimentation.



# Environment Setup





# Phishpedia Overview

## Key Features:

### 1. Logo Detection and Comparison:

- a) Phishpedia identifies logos on websites and matches them to known logos from a reference dataset. This helps determine if a website is impersonating a legitimate brand.

### 2. Uses Faster-RCNN for Logo Detection:

- a) Faster-RCNN (Faster Region-Based Convolutional Neural Network) is a deep learning model used for object detection.
- b) It scans the webpage screenshot and identifies areas where logos are likely present.

### 3. Siamese Network for Brand Logo Matching:

- a) After a logo is detected, the Siamese network compares it with known brand logos to see if it matches any legitimate brands.
- b) A Siamese network works by analyzing the similarity between two inputs, making it suitable for identifying subtle differences in logos.



# Phishpedia Overview

## Core Components:

### Scripts for Logo Recognition and Matching:

***logo\_recog.py***: This script is responsible for detecting logos on a webpage. It uses the Faster-RCNN model to locate logos in screenshots.

***logo\_matching.py***: Once a logo is detected, this script compares it with logos in the reference dataset using the Siamese network.

### Configuration File (*configs.yaml*):

- This file defines important parameters for the models, such as:
  - The detection threshold (minimum confidence required for logo detection).
  - The matching threshold (confidence level required to declare a match).
- Fine-tuning these thresholds improves the accuracy of detection and reduces false positives.



# Traditional Farmwork vs. Phishpedia

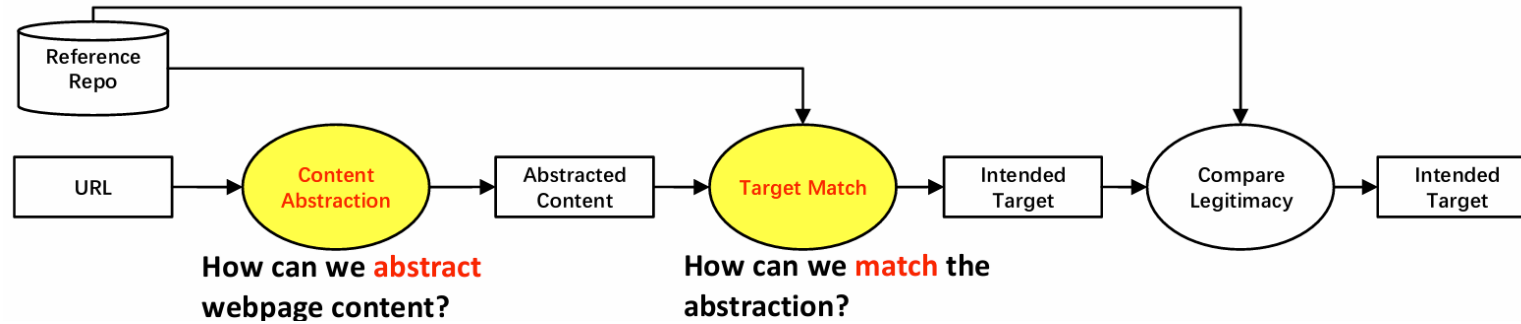


Fig1: Typical Framework for Phishing Identification

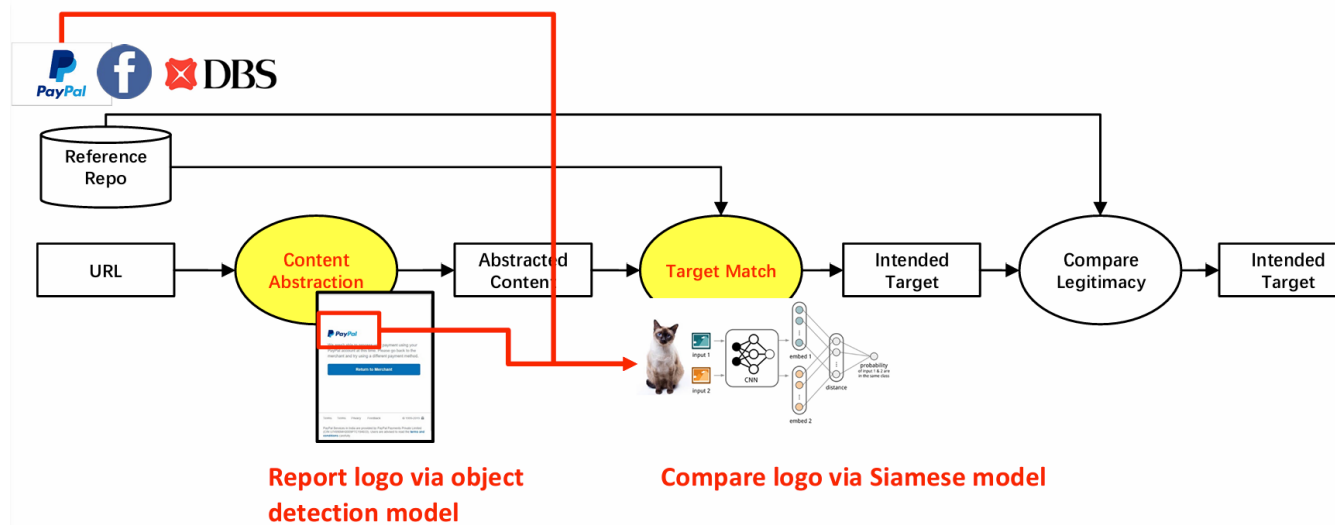


Fig2: Phishpedia Approach Overview





# Example of Phishing Site

✚ Let's go to PhishTank and find some malicious websites reported as phishing. Here is an example where we can see that it has been reported as phishing, and the prediction score is 100%.

The screenshot displays the PhishTank interface. At the top, the PhishTank logo is visible with the tagline "Out of the Net, into the Tank." Below the logo, a navigation bar contains links for Home, Add A Phish, Verify A Phish, Phish Search, Stats, FAQ, Developers, Mailing Lists, and My Account. The main content area shows a submission status: "Submission #8868696 is currently ONLINE". Below this, it states "Submitted Nov 25th 2024 7:00 AM by [@kashkashkash](#) (Correction: Nov 25th 2024 7:12 AM UTC)" and provides the URL "https://www.kashkashkashkash.my.id/sermek5/". A red banner at the bottom of the submission details reads "Verified: Is a phish" and "As verified by [Dev Jackson Stearns](#)". Below the banner, a progress bar shows "Is a phish" at 100% and "Is NOT a phish" at 0%.



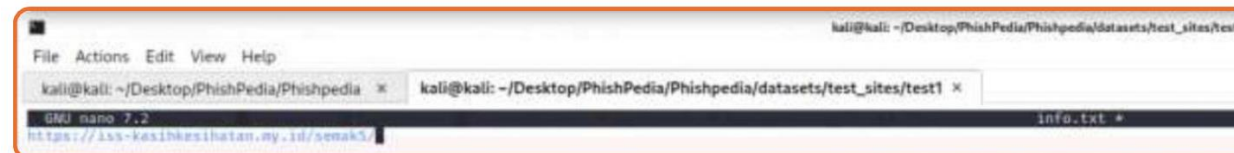
# Example of Phishing Site

🔗 Open in new tab



🔗 Create file named "info.txt" and add link to it

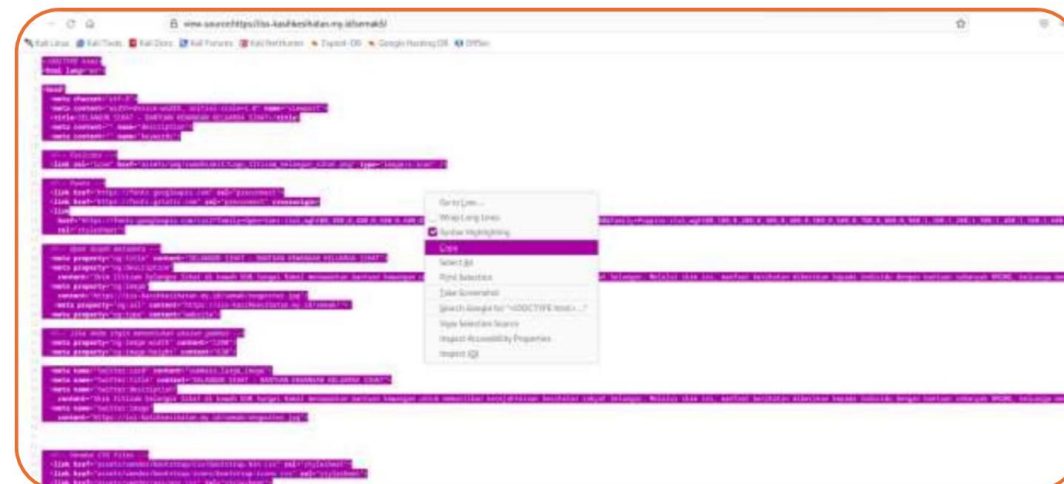
```
(phishpedia) (kali@kali) - [~/.../PhishPedia/Phishpedia/datasets/test_sites]  
$ nano info.txt
```





# Example of Phishing Site

- Take a screenshot of the webpage and add it to the test1 folder.
- Next, create a file named html.txt and paste the page source into it.
- Right-click on the webpage, select "View Page Source," and copy the content into the file.





# Example of Phishing Site

Now, we have all three required files

```
(phishpedia) (kali@kali) - [~/.../Phishpedia/datasets/test_sites/test1]
$ ls
html.txt  info.txt  snap.png
```

Now, let's run the tool

```
(phishpedia) (kali@kali) - [~/Desktop/PhishPedia/Phishpedia]
$ python3 phishpedia.py --folder ./datasets/test_sites
```

Reported as 100% phishing

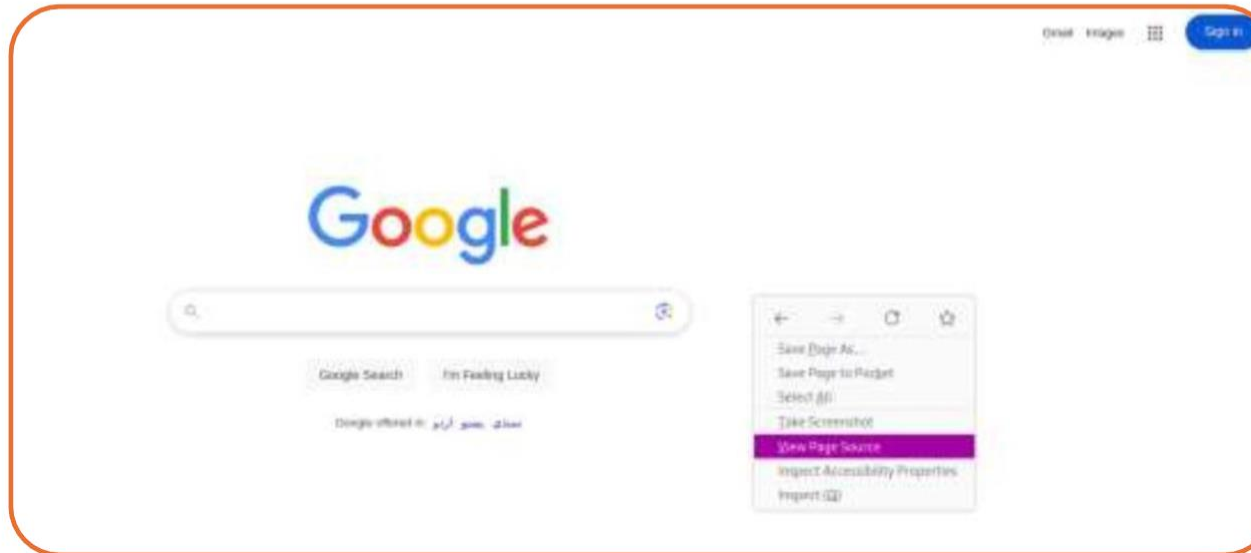
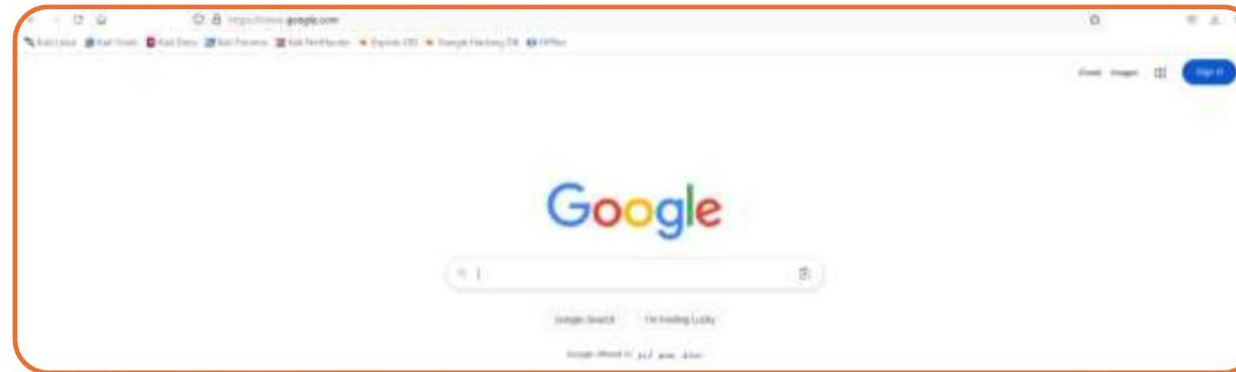
```
(phishpedia) (kali@kali) - [~/Desktop/PhishPedia/Phishpedia]
$ python3 phishpedia.py --folder ./datasets/test_sites
The checkpoint state_dir contains keys that are not used by the model:
state_dir
2021-04-04
[over protected logs list]
length of inference list = 2000
100%
```

Prediction: True



# Example of Legitimate Site

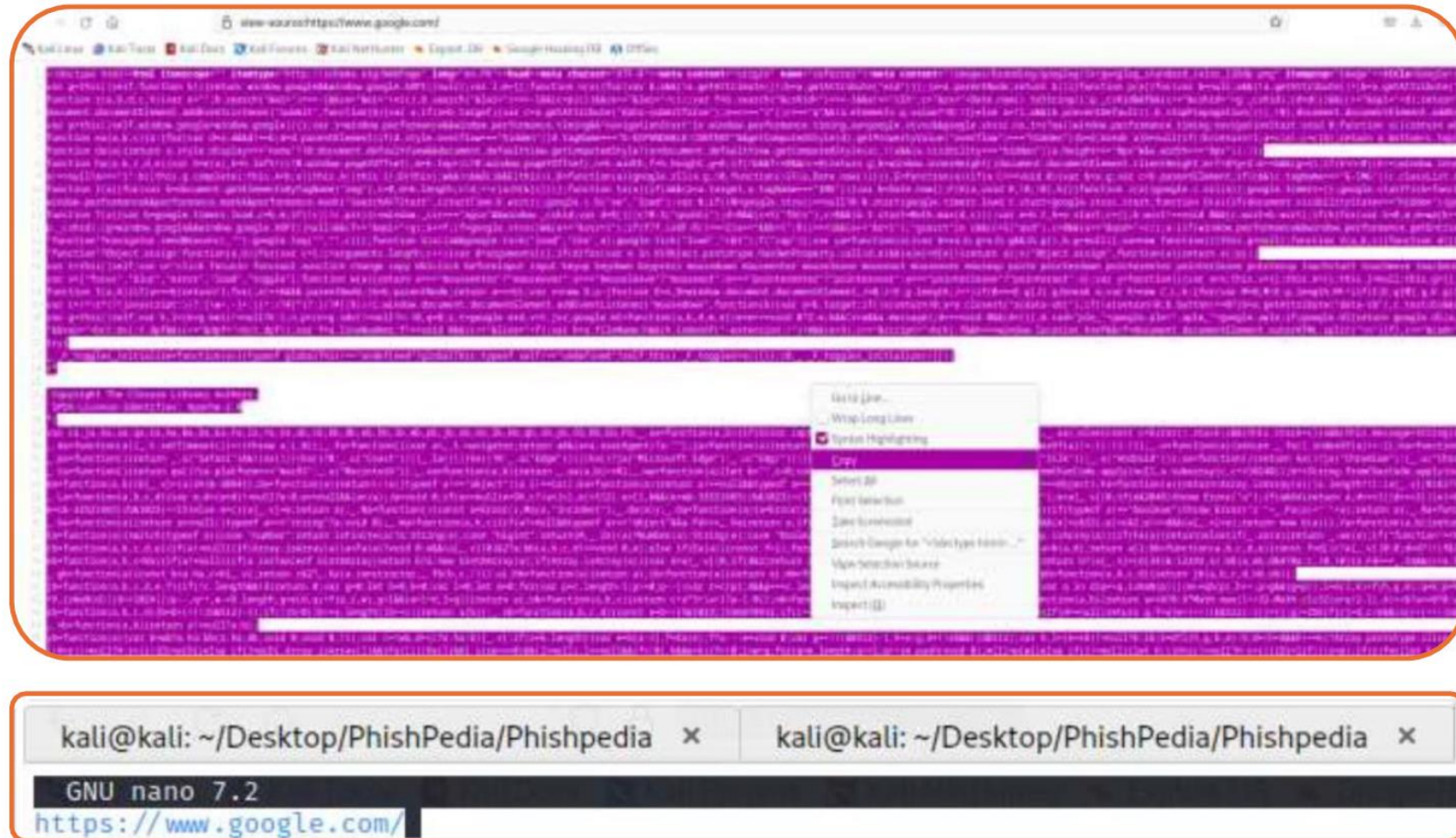
## 2.3.4.1. Site: Google.com







# Example of Legitimate Site





# Example of Legitimate Site

```
(base) (kali@kali)-[~/.../Phishpedia/datasets/test_sites/test2]  
└─$ ls  
html.txt  info.txt  snap.png
```

- ✚ Let's run the tool now
- ✚ Reported as 20% phishing

```
(phishpedia) (kali@kali)-[~/Desktop/PhishPedia/Phishpedia]  
└─$ python3 phishpedia.py --folder ./datasets/test_sites  
The checkpoint state_dict contains keys that are not used by the model:  
  pixel_mean  
  pixel_std  
Load protected logo list  
Length of reference list = 2996  
20%|██████████|
```

- ✚ Prediction: True



# Challenges and Limitations

- **Accuracy Concerns**

False positives when detecting logos due to low detection thresholds.

- **Scalability Issues**

Performance degradation with a growing brand reference list.

- **Dynamic Content Handling**

Difficulty analyzing dynamically loaded webpage elements (e.g., JavaScript-rendered content).

- **Adaptation to Evolving Threats**

Phishing tactics frequently evolve, requiring continuous updates to logo references and detection thresholds.

- **Environmental Constraints**

Dependency on pre-trained models and large datasets for accurate detection.





# Conclusion

learning and visual analysis to identify malicious websites with impressive accuracy. By focusing on logo detection and comparisons against trusted brand references, it provides a robust solution for identifying phishing attempts. Its reliability across a range of scenarios and domains underscores its potential as a critical component in modern cybersecurity frameworks. However, its limitations in scalability, detection of evolving phishing tactics, and handling dynamic content point to areas that require further development.



# Future work

- **Enhancing Scalability**

Investigate ways to optimize performance for larger brand reference datasets to handle real-world scaling challenges.

- **Improving Detection of Evolving Threats**

Update algorithms to adapt to rapidly changing phishing strategies, including obfuscated content and dynamic elements.

- **User Behavior Analysis:**

Incorporate models that analyze user interaction patterns on webpages to detect phishing attempts, such as unusual clicks or form submissions.

- **Integration with Real-Time Systems**

Develop seamless integration with browsers and email clients for live phishing detection.



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# References

Lindsey98. *Phishpedia: Phishing Detection Tool Using Deep Learning*.  
GitHub. Available at: <https://github.com/lindsey98/Phishpedia>.

Kali Linux, Anaconda, and PyTorch documentation.