Spot the Fake: Al-Powered Detection of Fraudulent Websites, Apps & Digital Content



# Al-Powered Detection of Fraudulent Websites & Digital Content

Cipher Cop 2025

Team Lead: Harsh Jain

Team Members: Nikhil Singh, Rishiraj Gupta, Sumit Kothari

Proactive Protection Through Multi-Modal Al Analysis

# The Digital Minefield

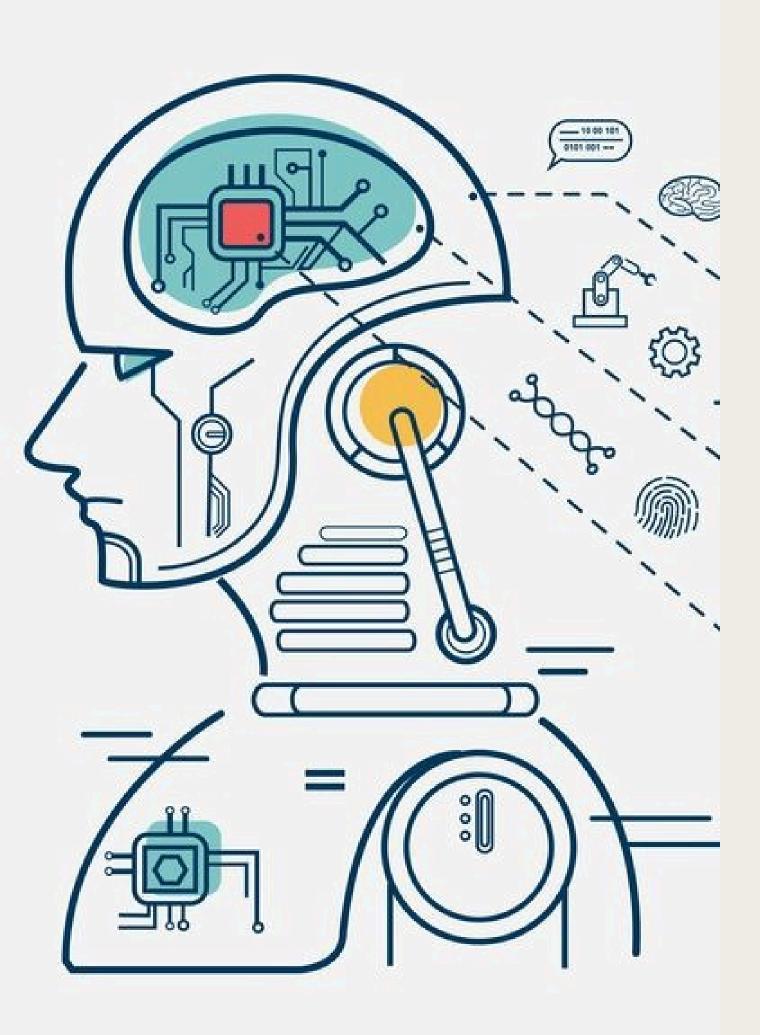
# The Digital Epidemic:

- Explosion of sophisticated fake websites and apps cloning trusted brands.
- Used for phishing, data theft, financial scams, and malware distribution.

# Why Current Solutions Fall Short:

- Reactive, not Proactive: Reliance on user reports and manual takedowns.
- Evolving Threats: Scammers constantly adapt, making simple rule-based systems ineffective.

Our Mission: To build a proactive, intelligent system that identifies fraud before users become victims.



## Our Solution - A Multi-Layered Al Shield

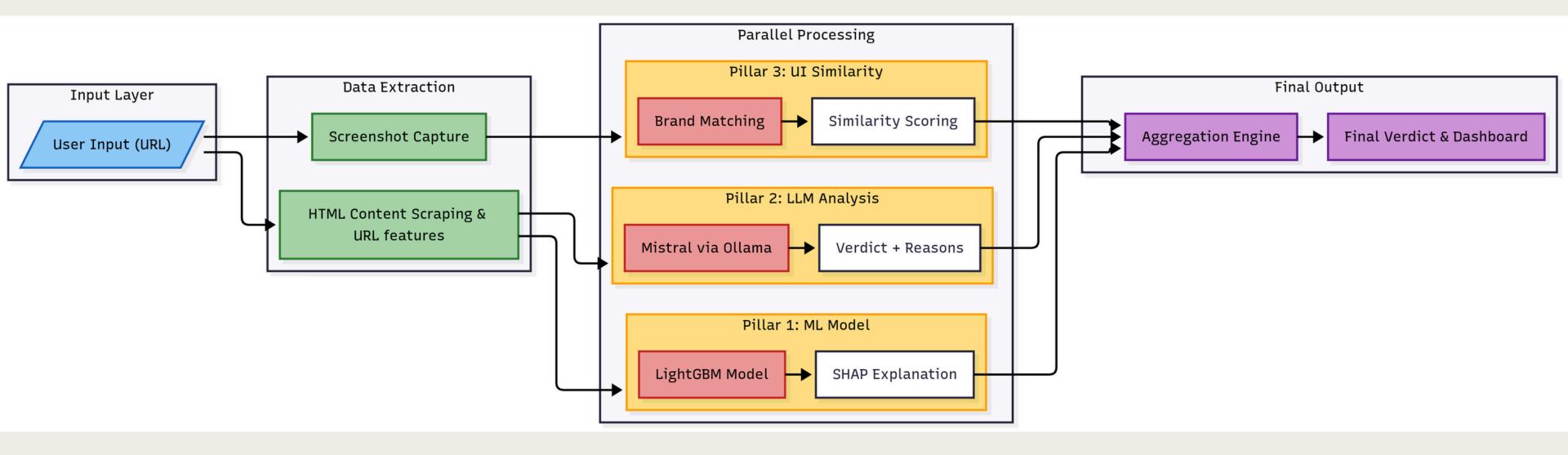
We developed a holistic defense system that analyzes multiple dimensions of digital content.

## **Three Pillars of Detection:**

- 1. Machine Learning (LGBM): Analyzes URL structure and features for classic phishing hallmarks.
- 1. Large Language Model (Mistral): Understands the *content* and context of the webpage for semantic analysis.
- 1. **UI Similarity Engine:** Uses Computer Vision to detect visual mimicry of known brands.

**Key Innovation:** Combining these three approaches into a single, weighted scoring system for superior accuracy and explainability.

Slide 4: Technical Architecture Overview



**Tech Stack:** Python, Streamlit, LightGBM, Ollama (Mistral), Selenium, OpenCV, pytesseract, SHAP, BeautifulSoup.

## Deep Dive 1 - The ML Engine (LightGBM)



#### What it does:

Analyzes over 30 heuristic features extracted directly from the URL.



#### **Features Include:**

- Length of URL/hostname, count of special characters (@, -, ~, %)
- Presence of IP addresses, punycode, or known shortening services.
- "Phishy" keywords (e.g., login, verify, account).
- Suspicious TLDs.



# Strength:

Extremely fast and great at catching obvious phishing patterns.

#### **Output:**

A probability score and a **SHAP explanation** showing *which features* most influenced the decision (e.g., "nb\_dots=10 was a strong phishing indicator").

#### Deep Dive 2 - The LLM Analyst (Mistral)



#### What it does:

Acts as a cybersecurity expert reading the page's content.

# SMS

#### **Process:**

Strength:

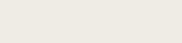
- 1. Extracts and cleans main text from the HTML.
- 1. Sends the content to a locally-run Mistral model via Ollama.
- 1. Forces a structured JSON response with a verdict, risk level, and, crucially, a list of **evidence snippets**.

양

## Sample LLM Output:

{"verdict": "phishing", "risk\_level": "suspicious",

"evidence\_snippets": \["Your account will be suspended' urgency
trigger", "Mismatch between domain 'secure-paypal-update.com'
and branded content"\]}



Understands nuance, social engineering tactics, and contextual clues that URL analysis misses.

#### **Deep Dive 3 - The UI Similarity Detective**



#### What it does:

Answers the question "Does this *look* like a real PayPal (or other brand) site?"



#### **Process:**

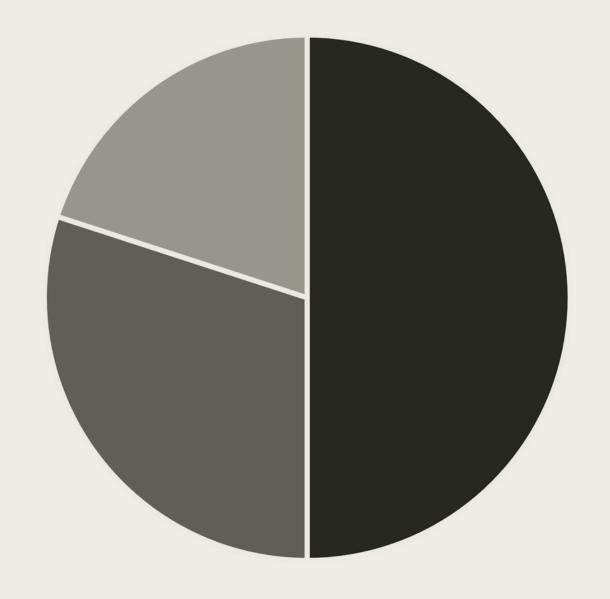
- 1. Takes a screenshot of the target site.
- Extracts the domain and performs fuzzy matching against a library of brand reference images (brand\_ref.png).
- Computes a composite similarity score based on: Perceptual Hash (pHash):
   Layout and structural similarity. Color
   Histogram: Color scheme and palette matching. \* OCR + Text Similarity: Text content and its stylistic presentation.



### Strength:

Catchs sophisticated visual clones that might bypass other checks.

## **The Ensemble Decision - Smarter Together**



## Our Weighted Scoring Algorithm:

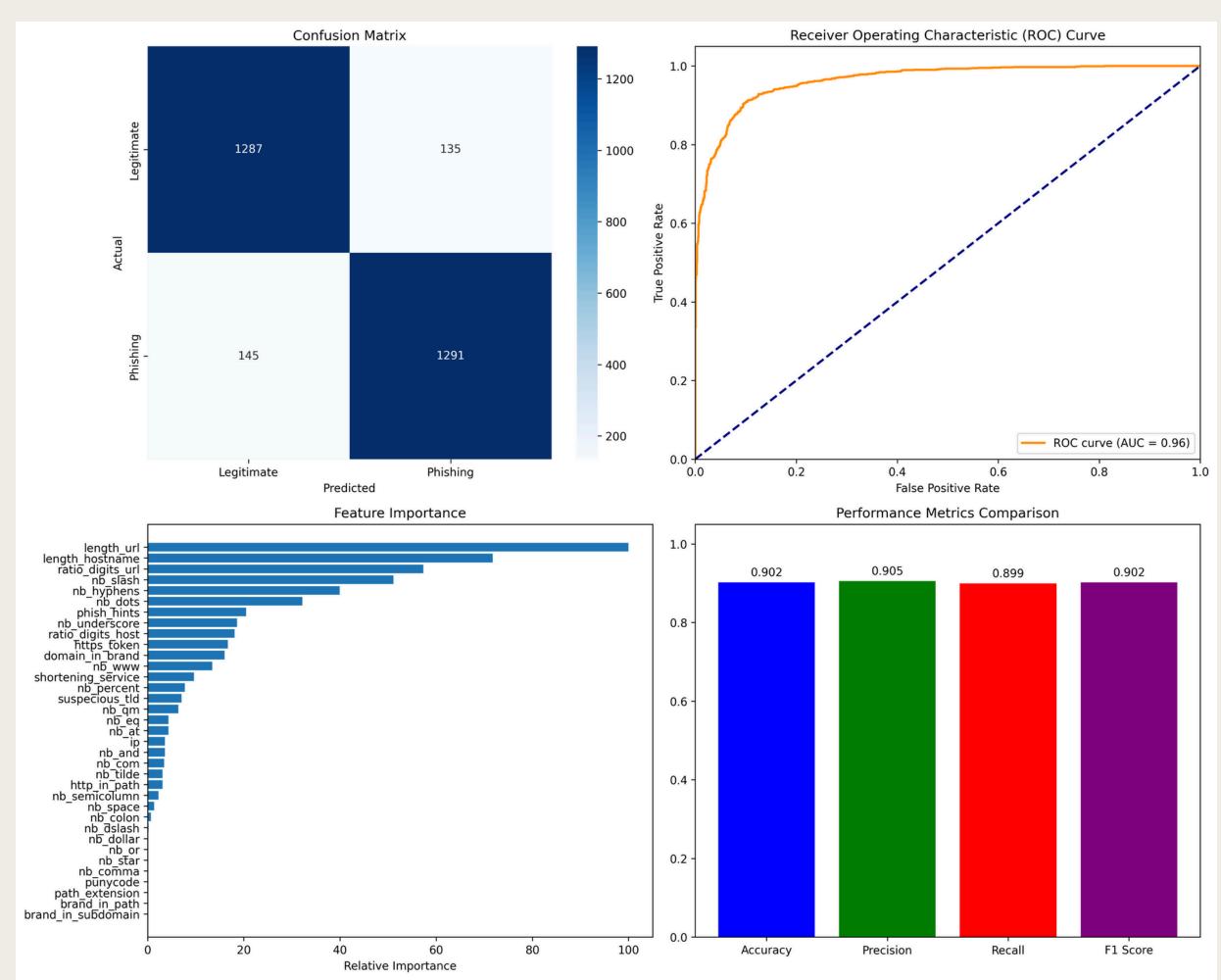
- LGBM Score (50% Weight): The foundational, quantitative check.
- LLM Score (30% Weight): The qualitative, contextual check.
- Ul Similarity Score (20% Weight): The visual authenticity check.

The system aggregates these into a **Final Legitimacy Score**.

# Why Ensemble?

No single method is perfect. Combining them reduces false positives/negatives.

# **LightGBM Performance Metrics**



### **Results & Impact**



#### **Successfully Created:**

A proactive, multi-modal detection system that meets the hackathon's objectives.



## **Key Achievements:**

- Explainable AI: Every decision is supported by clear evidence from each module.
- Holistic Approach: Combines structural, semantic, and visual analysis for comprehensive coverage.
- **Prototype Ready:** A user-friendly web app that could be extended into a browser extension.



#### **Potential Impact:**

Protects users from financial and data loss, reduces the effectiveness of phishing campaigns, and increases trust in digital platforms.