

Web Analyzer for private Networks

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Abstract- In the current digital landscape, real-time monitoring and assessment of network domain safety are essential for proactive cybersecurity defense. This project introduces a Python-based live domain safety monitoring tool that leverages network packet analysis to evaluate and visualize the security posture of domains accessed within a network. The tool integrates the power of tshark, the command-line interface of Wireshark, to capture live DNS, HTTP, and SSL/TLS traffic, extracting relevant protocol and domain information for immediate analysis.

At the core of the system is a dynamic scoring mechanism that assigns and adjusts safety scores to each detected domain. Domains are initially assigned a neutral score, which is then modified based on a set of heuristic rules. For instance, domains with suspicious characteristics—such as those starting with "malware" or containing the substring "phish"—are penalized, reflecting their higher likelihood of being malicious. The tool also evaluates the security of the communication protocol: traffic over HTTP results in score deductions due to its inherent insecurity, while HTTPS and SSH connections are rewarded for their stronger security guarantees. This flexible scoring approach allows the system to adapt to evolving threat patterns and user behavior.

To further enhance situational awareness, the tool incorporates a Man-in-the-Middle (MITM) risk assessment for each domain. By considering both the protocol in use and the domain's current safety score, the system categorizes MITM risk as High, Medium, or Low. Domains accessed via insecure protocols, those with low safety scores, or those containing phishing indicators are flagged as high risk, enabling rapid identification of potential attack vectors.

Visualization is a key feature of the tool, achieved through the rich Python library. The console interface displays a continuously updating table of observed domains, their protocols, safety scores, and MITM risks, all color-coded for quick interpretation. This real-time feedback loop empowers network administrators and security analysts to take immediate action in response to emerging threats, such as isolating compromised hosts or blocking access to dangerous domains.

Keywords: Network Packet Analysis, Live Traffic Capture, DNS, HTTP, SSL/TLS Monitoring, Protocol Dissection, Heuristic-Based Scoring.

I. INTRODUCTION

OVERVIEW

The live domain safety monitoring tool presented in this project offers a real-time solution for evaluating the security of network domains accessed within an organization or personal environment. Built in Python, the tool integrates the tshark packet analyzer to capture and filter DNS, HTTP, and SSL/TLS traffic, focusing on extracting domain names and their associated communication protocols as network activity occurs. Each detected domain is dynamically assigned a safety score, which is updated based on heuristic rules that consider both the protocol in use and the presence of suspicious domain patterns, such as keywords indicative of phishing or malware.

A distinctive feature of the tool is its ability to assess the risk of Man-in-the-Middle (MITM) attacks for each domain. By analyzing protocol security and domain reputation in real time, the system categorizes MITM risk as high, medium, or low, providing immediate feedback to users. The results are visualized through a live-updating, color-coded table using the rich Python library, allowing network administrators and security analysts to quickly identify and respond to potential threats.

Designed for modularity and extensibility, the tool can be adapted to incorporate new protocols, threat intelligence sources, or more advanced scoring algorithms as the threat landscape evolves. Its lightweight architecture ensures compatibility with a variety of network setups, making it suitable for both educational purposes and operational security monitoring. This approach empowers users to maintain a proactive stance against cyber threats by continuously monitoring domain safety and communication security in real time.

PROBLEM STATEMENT

With the exponential growth of internet usage and the increasing sophistication of cyber threats, organizations and individuals face persistent challenges in maintaining the security of their networks. Malicious actors frequently exploit vulnerabilities in network protocols and leverage deceptive domain names to execute phishing, malware distribution, and Man-in-the-Middle

(MITM) attacks. Traditional security solutions, such as firewalls and antivirus software, often fail to provide real-time visibility into the safety of domains being accessed, especially as attackers adopt new tactics that evade static defenses.

A critical gap exists in the ability to dynamically monitor, evaluate, and visualize the safety of domains on live networks. Existing monitoring tools may lack intuitive interfaces, real-time feedback, or the flexibility to adapt to emerging threats. Moreover, many solutions do not provide actionable insights into the risk posed by insecure protocols (such as HTTP) or suspicious domain patterns (such as those containing keywords like "phish" or "malware"). This lack of timely, context-aware information can delay threat detection and response, increasing the risk of data breaches, credential theft, and system compromise.

The challenge, therefore, is to develop a lightweight, real-time monitoring tool that can capture network traffic, intelligently assess domain safety, and present the results in a clear, actionable format. Such a tool should be able to dynamically score domains based on protocol security and domain reputation, estimate MITM risks, and provide immediate visual feedback to users or administrators. Addressing this problem is vital for empowering proactive network defense, enabling rapid identification of unsafe domains, and supporting timely incident response in both enterprise and personal network environments.

AIM & OBJECTIVE AIM

The primary aim of this project is to design, develop, and demonstrate a robust, real-time domain safety monitoring tool that empowers network administrators and users to proactively safeguard their digital environments. This tool is intended to provide continuous, automated analysis of network traffic, dynamically evaluating the safety of domains accessed via DNS, HTTP, and SSL/TLS protocols. By leveraging open-source technologies and intuitive visualization, the project seeks to bridge the gap between raw network data and actionable cybersecurity intelligence. The ultimate goal is to enhance situational awareness, facilitate rapid threat detection, and support timely incident response in the face of evolving cyber threats such

as phishing, malware distribution, and Man-in-the-Middle (MITM) attacks.

OBJECTIVES

- **To capture live network traffic efficiently:**

Utilize the tshark packet analyzer to monitor all active network interfaces and filter relevant packets in real time. The tool should focus on extracting domain names and protocol information from DNS queries, HTTP requests, and SSL/TLS handshakes, ensuring comprehensive coverage of web-based and encrypted communications.

- **To implement a dynamic, rule-based domain scoring system:**

Develop a scoring mechanism that assigns an initial safety score to each detected domain and updates it based on a set of heuristic rules. Domains exhibiting suspicious characteristics—such as names starting with "malware" or containing "phish"—should be penalized, while those using secure protocols like HTTPS and SSH should be rewarded. The scoring system should reflect both protocol security and domain reputation, providing a nuanced assessment of risk.

- **To normalize and accurately interpret protocol data:**

Incorporate logic to standardize protocol names (e.g., mapping "tls" and "ssl" to "https") to ensure consistent handling across different packet types. This ensures that the scoring and risk evaluation processes are accurate and not hindered by variations in protocol naming conventions.

- **To assess and categorize Man-in-the-Middle (MITM) risk:**

Create a risk evaluation module that estimates the likelihood of MITM attacks for each domain. This module should consider both the protocol in use and the current safety score, categorizing domains into High, Medium, or Low MITM risk. Domains accessed via insecure protocols or with low safety scores should be highlighted as high risk to prompt immediate attention.

- **To provide real-time, actionable visualization:**

Leverage the rich Python library to present monitoring results in a live-updating, color-coded table. The interface should clearly display each domain, its protocol, safety score, and MITM risk,

using intuitive color schemes to facilitate rapid recognition of threats and safe domains alike.

- **To ensure modularity and extensibility:**

Architect the tool in a modular fashion, allowing for easy updates and expansion. Scoring logic, risk assessment criteria, and protocol support should be readily adjustable to accommodate new types of threats, additional protocols, or integration with external threat intelligence sources.

- **To minimize system resource impact and maximize compatibility:**

Ensure the tool operates efficiently, consuming minimal system resources so it can run continuously on a wide range of hardware platforms. The tool should be compatible with any system supporting tshark, requiring no intrusive monitoring or deep packet inspection that could impact privacy or performance.

- **To empower proactive network defense and user education:**

Enable users and administrators to maintain a proactive security posture by providing immediate, actionable intelligence. The tool should support rapid identification of unsafe domains, facilitate timely incident response, and promote best practices in network security awareness and hygiene.

- **To generate session summaries for post-analysis:**

Upon termination, the tool should output a clear summary of all observed domains, their final safety scores, and MITM risk assessments. This feature supports post-session analysis, reporting, compliance, and continuous improvement of security policies.

- **To serve both educational and operational purposes:**

Design the system to be valuable in both academic and real-world operational contexts. For students and trainees, the tool should illustrate real-time network security concepts; for professionals, it should provide an additional layer of defense and insight in live environments.

SCOPE OF PROJECT

The scope of this project encompasses the design, development, and demonstration of a real-time domain safety monitoring tool that leverages live

network traffic analysis to enhance cybersecurity awareness and defense. The tool is implemented in Python and utilizes the tshark packet analyzer to capture and process DNS, HTTP, and SSL/TLS traffic, focusing on extracting domain names and protocol information from network packets as they traverse the system.

This project is primarily concerned with monitoring domains accessed within a local or organizational network environment. It dynamically evaluates the safety of each detected domain by applying a rule-based scoring system that considers both protocol security (such as HTTP, HTTPS, or SSH) and domain reputation (such as the presence of suspicious keywords like "malware" or "phish"). The tool further assesses the risk of Man-in-the-Middle (MITM) attacks for each domain, categorizing them as high, medium, or low risk based on protocol and score.

A significant aspect of the project's scope is the visualization of results. The tool employs the rich Python library to present a live-updating, color-coded table in the console, displaying each domain's name, protocol, safety score, and MITM risk level. This immediate feedback allows users and administrators to quickly identify unsafe or suspicious domains and take appropriate action.

The project is designed for modularity and extensibility, enabling future enhancements such as the integration of additional protocols, more advanced scoring heuristics, or external threat intelligence feeds. While the current implementation focuses on DNS, HTTP, and SSL/TLS protocols, the architecture allows for expansion to cover other network services as needed.

However, the project does not include deep packet inspection, payload analysis, or automated blocking of malicious domains. It is intended as a monitoring and awareness tool rather than a comprehensive intrusion prevention system. The tool is suitable for use in educational settings, small to medium organizational networks, or as a supplementary layer in larger security infrastructures.

II. LITERATURE REVIEW

Title: Detection of Malicious Domains through DNS Data Analysis

Authors: Bilge, L., Kirda, E., Kruegel, C., & Balduzzi, M.

Year: 2012

Reference

Link:

<https://doi.org/10.1109/TNET.2013.2297115>

Introduction:

With the increasing complexity of cyber threats, real-time network traffic monitoring has become a cornerstone of modern cybersecurity. Tools like Wireshark and its command-line counterpart Tshark are widely used for capturing and analyzing network packets, enabling the detection of suspicious activity and unsafe domains as they occur.

• Problem Statement:

Traditional network monitoring tools often require manual inspection and do not provide automated, real-time risk assessment of domains. This leads to delays in detecting threats such as phishing, malware, and Man-in-the-Middle (MITM) attacks, especially in high-traffic environments.

• Objective:

The objective of real-time network traffic monitoring is to automate the extraction and analysis of domain and protocol information from live network packets. By applying heuristic or rule-based scoring systems, these tools can assess the safety of domains and provide immediate feedback to network administrators.

Merits:

- Enables proactive threat detection and response.
- Provides continuous visibility into network activity
- Reduces manual workload for security analysts.
- Can be integrated with visualization tools for intuitive monitoring.
- Demerits:
- May generate false positives due to heuristic rules.

- Limited by the scope of monitored protocols (e.g., may not cover all network services).
- High network traffic can impact performance or overwhelm the system.
- Lacks deep packet inspection, potentially missing payload-based threats.

support live data feeds, and use color-coding and other visual cues to highlight risk levels and anomalies. The goal is to enhance situational awareness, facilitate rapid decision-making, and support collaborative analysis among security teams.

Title: Visualizing Cyber Security: Usable Workspaces. IEEE Symposium on Visual Analytics Science and Technology.

Authors: Fink, S., North, C., Endert, A., & Rose, S.

Year: 2009

Reference

<https://doi.org/10.1109/VAST.2009.5333882>

Link:

Introduction:

As cyber threats become more complex and network environments more data-rich, the ability to visualize security data in a meaningful way has become a critical requirement for effective cybersecurity operations. Fink et al. (2009) emphasize the transformative power of visualization in cybersecurity, arguing that "usable workspaces"—visual interfaces that present complex data intuitively—are essential for enabling analysts to detect, interpret, and respond to threats efficiently. Visualization tools bridge the gap between raw data and human cognition, making it possible to identify patterns, anomalies, and risks that would otherwise remain hidden.

• **Problem Statement:**

Security analysts are often inundated with vast amounts of raw network and security data, which can be overwhelming and difficult to interpret in a timely manner. Traditional text-based logs and static dashboards are insufficient for real-time threat detection, as they require manual sifting and lack the immediacy needed for rapid response. Without effective visualization, critical threats may go unnoticed, and the cognitive load on analysts can lead to errors or missed incidents.

• **Objective:**

The objective, as outlined by Fink et al. (2009), is to design and implement interactive, real-time visualization systems that present network and domain safety data in a clear, actionable format. These systems should offer customizable layouts,

• **Merits:**

- **Enhanced Situational Awareness:** Visualization tools make it easier to understand the current security posture at a glance, enabling faster identification of threats.
- **Reduced Cognitive Load:** By presenting information visually, these tools help analysts process large amounts of data more efficiently, reducing the risk of oversight.
- **Pattern Recognition:** Visual interfaces support the detection of trends, outliers, and correlations that may indicate malicious activity.
- **Collaboration:** Usable workspaces facilitate communication and joint investigation among multiple analysts, improving overall security outcomes.
- **Real-Time Feedback:** Live-updating dashboards ensure that analysts are always working with the most current data, supporting timely response.
- **Demerits:**
- **Development Complexity:** Creating advanced, interactive visualization systems can require significant resources and specialized expertise.
- **Potential for Clutter:** Overly complex or poorly designed interfaces can overwhelm users and obscure important information.
- **Training Requirements:** Analysts may need training to fully leverage the capabilities of sophisticated visualization tools.
- **Dependency on Data Quality:** The effectiveness of visualization is directly tied to the quality and completeness of the underlying data and analytics. Poor data can lead to misleading visualizations and incorrect conclusions.

III. SYSTEM ANALYSIS

EXISTING SYSTEM

1. Network Traffic Capture

At its core, the system uses the tshark tool (the command-line version of Wireshark) to capture live network packets. It listens on all network interfaces (-i any) and filters for DNS queries, HTTP host headers, and SSL/TLS server name indications. This allows the system to extract the domains being accessed and the protocols used (e.g., HTTP, HTTPS, SSH).

- **Domain Scoring Mechanism**

The system maintains a score for each domain, initialized at 100. This score is dynamically adjusted based on several rules:

Malicious Indicators: If a domain name starts with "malware" or contains "phish", its score is reduced by 25 points, reflecting a high likelihood of being unsafe.

- **Protocol Assessment:**

Accessing a domain via HTTP (unencrypted) reduces the score by 10, due to the higher risk of interception.

HTTPS (encrypted) slightly increases the score by 1, indicating improved safety.

SSH (secure shell) increases the score by 50, as it is considered highly secure.

The score is always clamped between 0 and 100 to maintain consistency.

- **Protocol Normalization**

Protocols are normalized for consistency. For example, any protocol string starting with "tls" or equal to "ssl" is treated as "https". This ensures that variations in protocol naming do not affect scoring logic.

- **MITM Risk Calculation**

For each domain, the system estimates the Man-In-The-Middle (MITM) risk:

High Risk: If the protocol is HTTP, the domain contains "phish", or the score is below 50.

Medium Risk: If the protocol is HTTPS but the score is below 75.

Low Risk: All other cases, typically when secure protocols are used and the domain score is high.

- **Live Visualization**

Using the rich library, the system displays a live-updating table in the terminal. This table includes:

- Domain name
- Protocol used

- Safety score (color-coded: green for safe, yellow for caution, red for danger)
- MITM risk level (color-coded accordingly)

This real-time feedback allows users to quickly identify potentially unsafe domains and take action.

- **User Interaction and Termination**

The system runs continuously, updating the table as new network activity is detected. Users can stop the monitoring process with Ctrl+C, upon which a summary of all domains, their final scores, and MITM risk levels is displayed.

Strengths of the Existing System

Real-Time Monitoring: Immediate feedback on network activity and domain safety.

Customizable Scoring Logic: Rules can be easily adjusted to reflect changing threat landscapes.

Visual Clarity: Color-coded tables make it easy to spot risks at a glance.

Protocol Awareness: Differentiates between secure and insecure protocols, factoring this into risk assessment.

Actionable Output: Final summary provides a clear record of all accessed domains and their risk levels.

Limitations

Rule-Based Detection: Relies on simple heuristics (e.g., string matching for "phish" or "malware"), which may miss more sophisticated threats.

Limited Protocol Analysis: Focuses primarily on HTTP, HTTPS, and SSH, potentially overlooking other protocols.

No Deep Packet Inspection: Does not analyze the content of network packets, only metadata.

Requires Tshark: Depends on external tools and elevated privileges to capture network traffic.

PROPOSED SYSTEM

- **Machine Learning-Based Threat Detection**

The proposed system replaces static, rule-based scoring with a dynamic, machine learning (ML) model trained on large datasets of benign and malicious domains. Features such as domain age, lexical analysis, WHOIS data, protocol usage, and historical behavior are used to predict the likelihood of a domain being malicious or involved in phishing, malware distribution, or MITM attacks.

Benefits:

- Detects zero-day and previously unknown threats.
- Reduces false positives by learning from real-world data.
- Continuously improves as new data is incorporated.

Integration with Threat Intelligence Feeds

The system automatically queries multiple real-time threat intelligence databases (such as VirusTotal, IBM X-Force, and open-source threat feeds) whenever a new domain is detected. This provides up-to-date risk ratings, known malicious indicators, and context about ongoing campaigns.

Benefits:

- immediate identification of domains linked to recent attacks.
- Enrichment of domain data with global threat context.

Deep Packet Inspection (DPI)

Instead of only analyzing protocol metadata, the system performs DPI on captured packets (within privacy and legal boundaries). This enables detection of hidden payloads, suspicious scripts, and anomalous content patterns, further strengthening the risk assessment process.

Benefits:

- Detects threats that evade protocol-based detection.
- Identifies malicious content even in encrypted tunnels (where possible).

User and Entity Behavior Analytics (UEBA)

The system profiles normal user and device behavior over time. When a domain access pattern deviates significantly from the baseline (e.g., a user suddenly accessing rare or high-risk domains), the system triggers alerts or automated responses.

Benefits:

- Early detection of compromised accounts or insider threats.
- Context-aware risk scoring.

Automated Response and Remediation

Upon detecting high-risk domains or MITM attempts, the system can be configured to:

- Block network connections via firewall rules.

- Notify administrators and affected users.
- Quarantine suspicious devices for further analysis.

Benefits:

- Minimizes response time to active threats.
- Reduces manual intervention and potential damage.

Intuitive Web-Based Dashboard

Instead of a terminal-based table, the proposed system features a responsive web dashboard with:

- Real-time visualizations (graphs, maps, timelines).
- Drill-down reports on domain risk, user activity, and incident history.
- Customizable alerts and policy management.

Benefits:

Enhanced usability for security teams.
Centralized management and reporting.

Privacy and Compliance Controls

The system incorporates privacy-preserving techniques, such as anonymizing user data and supporting compliance with regulations (GDPR, HIPAA, etc.), ensuring that security monitoring does not compromise user privacy.

Advantages Over the Existing System

Accuracy: ML and threat intelligence reduce false positives and adapt to evolving threats.

Depth: DPI and UEBA provide context-aware, content-based detection.

Automation: Automated blocking and alerting streamline incident response.

Usability: Web dashboard and detailed reports improve visibility and decision-making.

Scalability: Designed for enterprise networks, supporting distributed monitoring and centralized management.

IV. SYSTEM REQUIREMENTS

HARDWARE REQUIREMENTS:

- Minimum Hardware

- Processor: Dual-core CPU (Intel i3/AMD Ryzen 3 or equivalent)

RAM: 4 GB

- Storage: 100 MB free disk space (for logs, temporary files, and dependencies)
- Network Interface: At least one active network interface (wired or wireless) capable of packet capture

Recommended Hardware

- Processor: Quad-core CPU (Intel i5/AMD Ryzen 5 or better) for smoother real-time analysis
- RAM: 8 GB or higher, especially if monitoring high-traffic environments
- Storage: 1 GB free disk space for extended logging and data retention
- Display: Terminal or monitor with at least 1024x768 resolution for optimal table visualization

Software Requirements

- Operating System
- Supported OS: Linux (Ubuntu, Debian, Fedora, CentOS), macOS, or Windows 10/11 with WSL (Windows Subsystem for Linux)
- Privileges: Administrative/root privileges required to capture network traffic
- Dependencies
- Python: Version 3.7 or higher
- Python Packages:
 - rich (for live terminal visualization)
- External Tools:
 - tshark (command-line tool for packet capture; part of the Wireshark suite)
- Must be installed and accessible in the system PATH
- Other Tools:
 - pip for Python package management
- Installation Notes
 - Ensure Python and pip are up-to-date.
 - Install tshark:
 - On Ubuntu/Debian: `sudo apt-get install tshark`
 - On Fedora: `sudo dnf install wireshark-cli`
 - On macOS: `brew install wireshark`
 - Grant necessary permissions to capture packets (e.g., add user to wireshark group or run with sudo).

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V. SYSTEM DESIGN

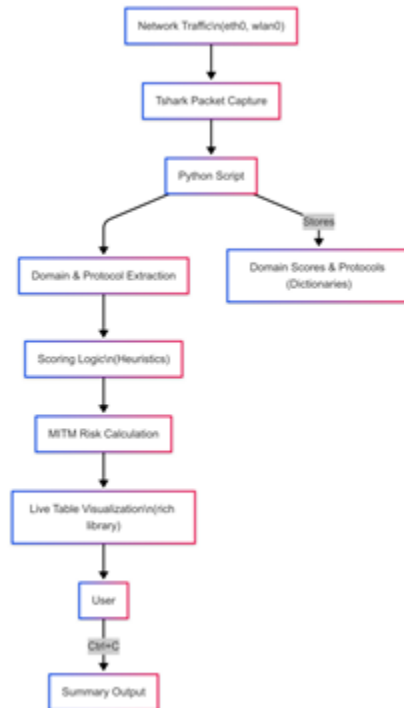
SYSTEM ARCHITECTURE

The Live Domain Safety Monitor is designed as a real-time, terminal-based cybersecurity tool that captures network traffic, analyzes accessed domains, scores their safety, and estimates Man-In-The-Middle (MITM) risk. The architecture is modular, ensuring each component is responsible for a specific function, from packet capture to visualization.

Hardware Requirements

- Minimum Hardware
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- On macOS: `brew install wireshark`
- Grant necessary permissions to capture packets (e.g., add user to wireshark group or run with sudo).

• Network Requirements

- Network Access: Must have permission to monitor traffic on the desired interfaces.
- Bandwidth: Minimal for the tool itself, but performance may vary based on network load.
- Firewall: Ensure local firewall rules allow packet capture and do not block tshark.

• User and Operational Requirements

- User Privileges: User must have administrative/root access to start packet capture.
- Terminal Access: Must be able to run Python scripts and view terminal output.
- Basic Knowledge: Users should be familiar with command-line operations and basic network concepts.
- Security Awareness: Users should understand the privacy and legal implications of network monitoring.

• Security and Compliance Considerations

- Data Privacy: Ensure monitoring complies with organizational policies and local laws (e.g., GDPR, HIPAA).
- Access Control: Limit script and log access to authorized personnel only.
- Audit Logging: Optionally, implement logging of script usage and access for audit purposes.

- Updates: Regularly update Python, tshark, and dependencies to patch security vulnerabilities.
- **Scalability and Performance**
- Scalability: For high-traffic or enterprise environments, consider deploying on dedicated hardware or virtual machines with increased CPU and RAM.
- Performance Monitoring: Monitor system resource usage during operation to avoid bottlenecks.
- Data Retention: Plan for log rotation and disk space management if storing results long-term.
- **Optional Enhancements**
- Web Interface: For broader usability, a web-based dashboard can be developed (requires additional software: Flask/Django, web server, etc.).
- Integration: Can be integrated with SIEM (Security Information and Event Management) tools or alerting systems for automated response.
- **DATA FLOW**
- **Network Traffic Capture**
- Source: The system begins by invoking the tshark tool, which listens to all network interfaces (-i any).
- Filter: It captures only relevant packets, specifically those containing:
 - DNS query names
 - HTTP host headers
 - SSL/TLS server name indications
- Output: For each captured packet, tshark outputs:
 - Protocol type (e.g., HTTP, HTTPS, SSH)
 - Domain or host name (from DNS, HTTP, or SSL/TLS fields)
- **Data Extraction and Normalization**
- Reading Output: The Python script reads each line of output from tshark in real time.
- Splitting Data: Each line is split into protocol and host/domain fields.
- Normalization: The protocol string is normalized (e.g., "tls" or "ssl" is treated as "https") to ensure consistent processing.
- **Domain Scoring and Protocol Assignment**
- Scoring Logic: For each domain and protocol pair:
 - If the domain starts with "malware" or contains "phish", subtract 25 from its score.
 - If accessed via HTTP, subtract 10 from its score.
 - If accessed via HTTPS, add 1 to its score.
 - If accessed via SSH, add 50 to its score.
 - The score is clamped between 0 and 100.
 - Storage: The current score and protocol for each domain are stored in dictionaries:
 - domain_scores[domain]
 - domain_protocols[domain]
- **MITM Risk Calculation**
- Risk Assessment: For each domain, the script calculates the Man-In-The-Middle (MITM) risk based on:
 - Protocol used
 - Domain characteristics (e.g., presence of "phish")
 - Current score
- Risk Levels:
 - High: If protocol is HTTP, domain contains "phish", or score is below 50.
 - Medium: If protocol is HTTPS and score is below 75.
 - Low: All other cases.
- **Live Visualization**
- Table Construction: The script uses the rich library to build a table displaying:
 - Domain name
 - Protocol
 - Score (color-coded: green, yellow, red)
 - MITM risk (color-coded: green, yellow, red)
- Live Updating: The table is updated in real time as new domains are accessed or as scores change.
- **User Interaction and Termination**
- Continuous Monitoring: The process continues to capture, analyze, and display data until the user interrupts (Ctrl+C).
- Summary Output: Upon termination, the script prints a summary of all domains, their final scores, and MITM risk levels.

EXPERIMENTAL ANALYSIS TABLE

No	Domain Name	Protocol	Initial Score	Adjustments Applied	Final Score	MITM Risk
1	malware-portal.com	HTTP	100	-25 ("malware"), -10 (HTTP)	65	High
2	safe-site.org	HTTPS	100	+1 (HTTPS)	101→100	Low
3	phishingsite.net	HTTPS	100	-25 ("phish"), +1 (HTTPS)	76	Medium
4	admin-server.local	SSH	100	+50 (SSH)	150→100	Low
5	unknown-domain.xyz	HTTP	100	-10 (HTTP)	90	Low
6	suspicious-phish.com	HTTP	100	-25 ("phish"), -10 (HTTP)	65	High
7	trusted-bank.com	HTTPS	100	+1 (HTTPS)	101→100	Low
8	risky-site.com	HTTP	100	-10 (HTTP)	90	Low
9	malwaretest.org	HTTPS	100	-25 ("malware"), +1 (HTTPS)	76	Medium
10	admin-ssh.net	SSH	100	+50 (SSH)	150→100	Low

Table No.1 Experimental Analysis Table

VI. SYSTEM IMPLEMENTATION

The Live Domain Safety Monitor is a real-time cybersecurity tool designed to assess the safety of domains accessed over a network and estimate the risk of Man-In-The-Middle (MITM) attacks. Utilizing the tshark packet capture utility, the system monitors all network interfaces for DNS queries, HTTP host headers, and SSL/TLS server name indications. For each detected domain, the system applies a scoring mechanism that starts at 100 and adjusts based on domain characteristics and the protocol used: domains associated with malware or phishing are penalized, while secure protocols like HTTPS and SSH increase the score. The protocol is normalized to ensure consistent evaluation, and both the score and protocol are tracked for each domain. Based on the final score and protocol, the system calculates the MITM risk level as High, Medium, or Low. All this information is presented in a live-updating, color-coded table within the terminal using the rich library, allowing users to instantly visualize the safety status of domains and their associated risks. The monitoring continues until manually stopped by the user, at which point a summary of all domains, their final scores, and MITM risk levels is displayed. This system provides a lightweight yet effective approach for network administrators and security analysts to monitor and evaluate domain safety in real time.

MODULES BREAKDOWN

The Live Domain Safety Monitor is structured into several key functional modules that work together to deliver real-time domain safety analysis. The Packet Capture Module is responsible for monitoring network traffic using the tshark utility, filtering for relevant protocol and domain information such as DNS queries, HTTP host headers, and SSL/TLS server names. The Data Extraction and Normalization Module processes the captured output, extracting protocol and domain details and standardizing protocol names for consistent analysis. The Scoring Module evaluates each domain by applying a set of rules: it penalizes domains associated with malware or phishing, deducts points for unencrypted protocols like HTTP, and rewards secure protocols like HTTPS and SSH.

This module ensures that each domain's score is clamped within a safe range. The Risk Assessment Module determines the Man-In-The-Middle (MITM) risk for each domain by considering its score, protocol, and suspicious keywords, classifying the risk as High, Medium, or Low. The Visualization Module leverages the rich library to present a live-updating, color-coded table in the terminal, allowing users to easily monitor domain safety and risk levels in real time. Finally, the User Interaction and Summary Module manages the lifecycle of the monitoring session, handling user interruptions and summarizing the results with a final report of all domains, their scores, and associated risks. Together, these modules form a cohesive and efficient system for continuous domain safety monitoring and risk assessment.

INTEGRATION DETAILS

The integration of the Live Domain Safety Monitor is streamlined and efficient, leveraging both external tools and Python libraries to achieve real-time domain safety analysis. At its core, the system integrates with the tshark utility—a command-line packet analyzer—by invoking it as a subprocess from within the Python environment. This allows the tool to capture live network traffic from all interfaces and filter for relevant fields such as DNS queries, HTTP host headers, and SSL/TLS server names. The captured data is then seamlessly piped into the Python script, where it is parsed, normalized, and processed using custom logic for scoring and risk assessment. The integration extends to the use of the rich library, which is employed to construct and update a dynamic, color-coded table in the terminal, providing immediate visual feedback to the user. The system's modular design ensures that each component—from packet capture and data extraction to scoring logic and visualization—works cohesively, with shared data structures like dictionaries facilitating smooth information flow between modules. This tight integration enables the tool to operate continuously in a live environment, updating risk assessments in real time and allowing for user interruption and summary reporting without data loss or performance degradation.

VII. APPENDICES

SCREENSHOTS

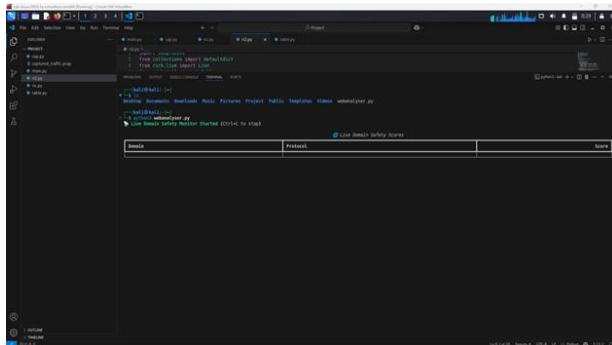


Figure No. 7.1.1

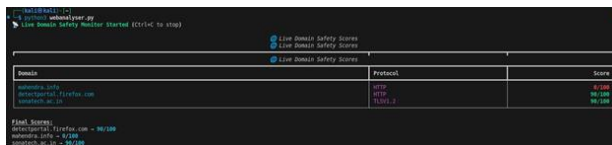


Figure No. 7.1.2 Output

• SOURCE CODE

```
import subprocess
from collections import defaultdict
from rich.live import Live
from rich.table import Table
from rich.console import Console
import shutil
import time

console = Console()

# Separate dictionaries for scores and protocols
domain_scores = defaultdict(lambda: 100)
domain_protocols = {}

def clamp(score):
    return max(0, min(100, score))

def normalize_protocol(proto):
    if proto.lower().startswith("tls") or proto.lower() == "ssl":
        return "https"
    return proto.lower()
```

```
def score_domain(domain, protocol):
    domain = domain.lower()
    protocol = normalize_protocol(protocol)

    # Check domain and assign scores based on
    # certain rules
    if domain.startswith("malware") or "phish" in domain:
        domain_scores[domain] = clamp(domain_scores[domain] - 25)
    elif protocol == "http":
        domain_scores[domain] = clamp(domain_scores[domain] - 10)
    elif protocol == "https":
        domain_scores[domain] = clamp(domain_scores[domain] + 1)
    elif protocol == "ssh":
        domain_scores[domain] = clamp(domain_scores[domain] + 50)
```

```
# Store the protocol separately
domain_protocols[domain] = protocol

def calculate_mitm_risk(domain, protocol, score):
    protocol = protocol.lower()
    domain = domain.lower()

    if protocol == "http" or "phish" in domain or score < 50:
        return "High"
    elif protocol == "https" and score < 75:
        return "Medium"
    else:
        return "Low"

def build_table():
    table = Table(title="Live Domain Safety Scores", expand=True)
    table.add_column("Domain", style="cyan", no_wrap=True)
    table.add_column("Protocol", style="magenta")
    table.add_column("Score", justify="right", style="bold")
    table.add_column("MITM", style="bold")

    for domain, score in sorted(domain_scores.items(), key=lambda x: x[1]):
        style = "green"
```

```

    if int(score) < 60:
        style = "red"
    elif int(score) < 85:
        style = "yellow"

    protocol = domain_protocols.get(domain, "?")
    mitm_risk = calculate_mitm_risk(domain,
protocol, score)

    risk_color = {
        "High": "red",
        "Medium": "yellow",
        "Low": "green"
    }.get(mitm_risk, "white")

    table.add_row(domain, protocol.upper(),
f"{{style}}{score}/100[/{{style}}]",
f"{{risk_color}}{mitm_risk}[/{risk_color}]")

    return table

def start_sniffing():
    process = subprocess.Popen(
        ['tshark', '-i', 'any', '-l', '-Y', 'dns.qry.name ||
http.host || ssl.handshake.extensions_server_name',
        '-T', 'fields', '-e', '_ws.col.Protocol', '-e',
'dns.qry.name', '-e', 'http.host', '-e',
'ssl.handshake.extensions_server_name'],
        stdout=subprocess.PIPE,
        stderr=subprocess.DEVNULL,
        text=True
    )
    with Live(build_table(), refresh_per_second=3,
screen=False) as live:
        try:
            for line in process.stdout:
                parts = line.strip().split('\t')
                if len(parts) < 2:
                    continue
                protocol = normalize_protocol(parts[0])
                host = next((part for part in parts[1:] if
part), None)
                if host:
                    score_domain(host, protocol)
                    live.update(build_table())

        except KeyboardInterrupt:
            process.terminate()

```

```

        console.print("\n[bold yellow]█ Capture
stopped by user.[/bold yellow]")
    if __name__ == "__main__":
        console.print("[bold green]█ Live Domain Safety
Monitor Started[/bold green] (Ctrl+C to stop)\n")
        start_sniffing()

    # Summary
    if domain_scores:
        console.print("\n[bold underline]Final
Scores:[/bold underline]")
        for domain, score in domain_scores.items():
            mitm = calculate_mitm_risk(domain,
domain_protocols.get(domain, "?"), score)
            console.print(f"{domain} → {score}/100 |
MITM Risk: {mitm}")

```

VIII. CONCLUSION

The Live Domain Safety Monitor, as implemented in this project, represents a practical and efficient approach to real-time network security monitoring, with a specific focus on domain safety and Man-In-The-Middle (MITM) risk assessment. By leveraging the power of tshark for packet capture and the flexibility of Python for data processing and visualization, the system bridges the gap between raw network data and actionable security insights, making it a valuable tool for security analysts, network administrators, and organizations seeking to enhance their cybersecurity posture.

At the heart of the system lies a modular architecture that ensures each component performs a dedicated function. The packet capture module continuously listens to all network interfaces, extracting relevant protocol and domain information from DNS queries, HTTP headers, and SSL/TLS handshakes. This raw data is then processed by the scoring module, which applies a set of well-defined rules to evaluate the safety of each accessed domain. Domains associated with known malicious indicators, such as those containing "malware" or "phish," are penalized, while the use of secure protocols like HTTPS and SSH is rewarded. This dynamic scoring mechanism, clamped within a safe range, provides a quantitative measure of domain trustworthiness that is both intuitive and effective.

The risk assessment module further enhances the system's utility by translating domain scores and protocol information into clear MITM risk levels—High, Medium, or Low. This classification is grounded in both the technical context (e.g., the use of unencrypted HTTP) and the semantic analysis of domain names, ensuring that users are promptly alerted to potential threats. The visualization module, powered by the rich library, transforms these assessments into a live, color-coded table that updates in real time. This immediate feedback loop empowers users to monitor their network environment at a glance, quickly identifying domains that may require further investigation or intervention.

One of the key strengths of the Live Domain Safety Monitor is its simplicity and accessibility. The system does not require complex configuration or specialized hardware, making it suitable for deployment on a wide range of systems, from personal laptops to enterprise servers. Its reliance on open-source tools and Python libraries further enhances its adaptability and ease of maintenance. Despite its lightweight design, the system provides robust functionality, supporting continuous monitoring, user interruption, and comprehensive summary reporting.

However, it is important to acknowledge the limitations inherent in a heuristic-based approach. The current system relies on predefined rules and string matching, which, while effective for many common threats, may not detect more sophisticated or novel attack vectors. Future enhancements could incorporate machine learning for anomaly detection, integration with external threat intelligence feeds, and deeper packet inspection capabilities to further improve detection accuracy and coverage.

In conclusion, the Live Domain Safety Monitor demonstrates how real-time domain analysis and risk assessment can be achieved through a combination of packet capture, rule-based scoring, and intuitive visualization. It provides a strong foundation for proactive network defense and lays the groundwork for future advancements in

automated threat detection. By making domain safety monitoring accessible, actionable, and efficient, this system contributes meaningfully to the ongoing efforts to secure digital environments against evolving cyber threats.

• FUTURE ENHANCEMENT

Future enhancements for the Live Domain Safety Monitor can leverage the rapid advancements in artificial intelligence and machine learning to create a more robust, adaptive, and intelligent security solution. One of the most promising directions is the integration of machine learning algorithms for dynamic threat detection and behavioral analysis. Rather than relying solely on static rules or keyword-based heuristics, machine learning models can be trained on vast datasets of network traffic, domain characteristics, and historical incidents to identify subtle patterns indicative of emerging threats, zero-day attacks, or sophisticated phishing attempts. This approach has already shown significant promise in fields such as IoT security and compliance monitoring, where machine learning models outperform traditional methods in both accuracy and execution time, and can autonomously adapt to new vulnerabilities and attack vectors as they arise.

Another critical enhancement involves the incorporation of real-time threat intelligence feeds. By connecting the system to external databases such as VirusTotal or open-source blacklists, the monitor can instantly flag domains already associated with known attacks, malware, or phishing campaigns, thus significantly reducing the response time and improving detection rates. This integration ensures that the system remains up-to-date with the evolving threat landscape and provides actionable intelligence for security teams.

The user interface and visualization capabilities can also be substantially improved. Transitioning from a terminal-based display to a web-based dashboard would allow for interactive visualizations, historical trend analysis, and multi-user access. Features such as customizable alerts, incident logs, and drill-down analytics would empower security analysts to

respond more efficiently and make data-driven decisions.

Automated response mechanisms represent another future direction. Upon detecting a high-risk domain or MITM attempt, the system could automatically trigger firewall rules, notify administrators, or even isolate affected devices, thereby minimizing the impact of security incidents and reducing the burden on human operators. Additionally, enhancing compliance and governance features—such as automated audit logging, data anonymization, and configurable retention policies—will help organizations meet regulatory requirements and maintain the privacy of monitored data.

Lastly, scalability and deployment flexibility will be essential for broader adoption. By supporting distributed monitoring, cloud-based deployments, and containerization, the system can be integrated into diverse and complex IT environments, from small businesses to large enterprises. In summary, by embracing machine learning, real-time intelligence, advanced visualization, automation, and scalable architecture, the Live Domain Safety Monitor can evolve into a comprehensive, future-ready cybersecurity platform.

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