

SRI RAMACHANDRA FACULTY OF ENGINEERING AND TECHNOLOGY

# CERTAIN INVESTIGATION OF MALICIOUS AND NORMAL TRAFFIC INTERPRETATION

INT 400 - INTERNSHIP

PROJECT REPORT

Submitted by

**AKISH RAJ A - E0222047** 

DHARSON RAM - E0222025

In partial fulfilment for the award of the degree of

**BACHELOR OF TECHNOLOGY** 

in

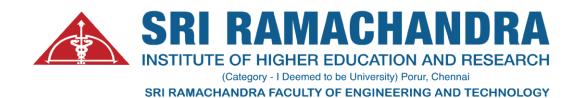
COMPUTER SCIENCE AND ENGINEERING

(Cyber Security & Internet of Things)

Sri Ramachandra Faculty of Engineering and Technology

Sri Ramachandra Institute of Higher Education and Research, Porur, Chennai - 600116

OCTOBER, 2024



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#### **BONAFIDE CERTIFICATE**

Certified that this project report "CERTAIN INVESTIGATION OF MALICIOUS AND NORMAL TRAFFIC INTERPREATION" is the bonafide record of work done by "AKISH RAJ A – E0222047" who carried out the internship work under my supervision.

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#### **ACKNOWLEDGEMENT**

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### TABLE OF CONTENTS

CHAPTER NO.	TITLE	PAGE NO.
	ABSTRACT	vii
	LIST OF TABLES	vi
1	INTRODUCTION	1
	1.1.Introduction	1
	1.2 Techniques Involved	1
	1.2.1 Packets capturing with Wireshark	1
	1.2.2 Filtering and Analysis	2
	1.2.3 PCAP File Analysis Using Python	2
	1.2.4 Automation with TShark	2
	1.3 Data Collection	2
2	LITERATURE REVIEW	5
3	PROPOSED METHODOLOGY	6
4	IMPLEMENTATION	6-7
5	RESULTS AND DISCUSSION	8-13
6	CONCLUSION AND FUTURE WORK	22
7	APPENDICES	
	Appendix-1: Code – Technical detail	15-21
	Appendix-2: Screenshots	8-13
8	REFERENCES	23
9	WORKLOG	24

### LIST FIGURES

Fig No	FIGURE NAME	PAGE NO	
1.	WORKFLOW	6	
2.	CAPTUARING NORMAL TRAFFIC	8	
3.	CAPATUARING	8	
	RETRANSMISSION ISSUE		
4.	RUNNING A WEBSITE TO	9	
	CAPTURE TRAFFIC		
5.	SCANING HTTP TRTAFFIC	9	
6.	ANALYSING CAPTURE DATA	10	
7.	STARTING MALWARE ARP	11	
	POISIONING		
8.	HOST IP ADDRESS	11	
9.	SCANING ARP USING	12	
	WIRESHARK		
10	NETWORK SCANING USING	12	
	TSHARK	_	
11	PCAP FILE SCANING	13	

#### **ABSTRACT**

The project Certain Investigation on Malicious and Normal Traffic Interpretation aimed to explore selected network problems by examining in the Wireshark traffic captured related with either malware, slow connection or other anomalies, etc Using various Wireshark filters, this project captures and analyzes the traffic to see different patterns in incoming data as normal versus malicious behavior. Here the Python is to process the packet capture (PCAP) files saved so far into a structured format destined for further analysis. The project consists of plotting IP address destinations on a map to visualize where traffic is going. The project also contains some automation scripts written in TShark (Wireshark's CLI version) to help with repetitive and simplified traffic analysis. Together, these build bottoms-up layer of rationale necessary to address the network security challenges in a holistic manner that includes automation at every level.

#### INTRODUCTION

#### 1.1. INTRODUCTION

Differentiating between regular and malicious traffic is required to protect against cyber-attacks, detect performance problems, and support stable operation. The project: Certain Investigation of Malicious and Normal Traffic Interpretation uses Wireshark, which is one among the various network protocol analyzers that allow logs to be viewed interactively or offline preprocess analysis on hosts in mannerable pattern for identification from malicious problems like malware and slow connections. Finally, the results are analyzed in python which takes PCAP files as input and based on destination IP provides good information on how traffic flows geographically. Also, TShark enables analyzer automation scripts of network traffic, it substitutes manual works and analyze processes greatly simplified. The sum of these parts provides a powerful source for monitoring and security investigations on your network in real time.

#### 1.2. TECHNIQUES INVOLVED

#### 1.2.1 Packet Capturing with Wireshark:

Wireshark is a network packet analyzer that enables you to capture and view packets traveling on a live network. This allows a detailed insight while looking into packets data for normal as well as abnormal activities. Variety: Various kinds of filters are added to segregate certain traffic such as Malware, Poor Connection and different out of the box patterns.

#### 1.2.2 Filtering and Analysis:

With the help of Wireshark filter options, you can easily sift through large amounts of traffic and look specifically at what protocol is having trouble. This allows for profiling of packet information such as source, destination IP addresses; protocols and flags to identify possible anomalies or security events.

#### 1.2.3 PCAP File Analysis Using Python:

There are pcap files that have been saved from Wireshark which need processing in Python. This data is then sniffed, pulled out via Python scripts (ZIP) and used to understand the network traffic by analyzing it. A major conclusion that can be drawn from this analysis is location of the destination IP geolocations to show on a map about where the traffic goes.

#### 1.2.4 Automation with TShark:

TShark is used for traffic analysis tasks from automated scripts and is the command line version of Wireshark in some way. TShark is also scripted in small automation scripts which can be used to capture network traffic, filter it and then analyze it without human interaction making the whole process automated.

#### 1.3. DATA COLLECTION

Wireshark is used to collect network traffic, which is saved in PCAP files. These filters are protocol-based or target a type of suspicious behavior It extracts key metadata such as source/destination IP, and timestamps for further analysis. Python scripts do this to convert the destination IPs into geolocations which it help visual mapping. With TShark being automated, you have an automatic traffic capture running in the background.

#### LITERATURE SURVEY

## 1.Title: Network Traffic Analysis Using Wireshark and TShark Automation

Patel and Kumar (2023) propose the use of network traffic analysis methods utilizing Wireshark and describe its usefulness in cases of malicious activities. According to the authors, the program has the real-time packet-filtering function, as well as joining to TShark. Authors of this work also refer to the role of Python in location and graphical representation of traffic patterns and emphasize how seeing an IP address plotted on a map helps detect the threat. The paper ends discussing the need and possibility of automating some aspects of the network traffic analysis for the sake of improving network security in different environments.

# 2.Title: Automation of Network Traffic Monitoring Using Python and TShark

Singh et al. (2022) explain the need for the integration of TShark into Python scripts in an automated manner in order to perform continuous operational traffic Analysis. While performing this duty, they stress the application of automation tools in order to minimize on human effort. The investigation presented in this research was intended to explore how automated scripts could enhance monitoring of traffic and threats without extensive contact with geolocation data and machine learning predictive models based on existing scripts.

#### 3. Title: Real-Time Network Traffic Analysis Using Machine Learning

Gomez et al. (2023) investigate the application of machine learning in real-time network traffic analysis. They have propose a model combining packet filtering and feature extraction techniques from traffic data using Wireshark and TShark. The authors demonstrate how automation tools improve detection of anomalies by integrating classification algorithms like Random Forest and SVM. Their research highlights the efficiency of using machine learning models for both normal and malicious traffic, emphasizing the increased detection accuracy with real-time monitoring.

#### 4.Title: Detecting Network Anomalies Through Traffic Patterns

Zhao and Fernandez (2022) focus on detecting anomalies by analyzing traffic patterns from Wireshark-collected data. Their research outlines the use of Python to automate geolocation mapping, which tracks the origin of suspicious IP addresses. They emphasize the benefits of combining packet analysis with geolocation visualization in identifying malicious traffic. Additionally, the study provides insights into implementing advanced algorithms to classify the nature of traffic flow, helping reduce false positives in detection.

## **5.Title: Visualizing Malicious Network Traffic Using Geolocation Techniques**

Iqbal and Rathi (2023) explore the use of geolocation mapping techniques to visualize malicious network traffic. The authors employ Wireshark to capture traffic and Python-based geolocation libraries to identify the origin of potential threats. Their study discusses how mapping can reveal geographic clusters of malicious activity, aiding cybersecurity efforts. The authors suggest combining machine learning techniques for deeper traffic analysis and identifying new attack vectors. The research is valuable for enhancing threat visualization capabilities.

#### 6.Title: Automation of Network Traffic Capture for Cybersecurity

Khan et al. (2024) present a framework that automates network traffic capture using TShark and Python for continuous monitoring. Their research demonstrates how automation reduces the manual effort required for monitoring large networks while improving detection of anomalies in real time. They discuss strategies to integrate Python-based scripts to automate traffic collection and analysis, highlighting the importance of scalability for larger networks. Their study underlines the effectiveness of automation in identifying emerging cyber threats.

## 7.Title: Enhancing Network Security with Wireshark and Python Automation

Ramos and Ali (2022) examine the application of Wireshark and Python for enhancing network security by automating packet analysis. They present case studies demonstrating the detection of abnormal traffic and how geolocation techniques in Python help trace attacks back to their sources. Their findings suggest that automation is key in handling the growing complexity of modern network environments, improving the accuracy and speed of detecting security breaches.

# **8.Title: Anomaly Detection in Network Traffic Through Automated Scripting**

Patel et al. (2024) evaluate the use of automated scripts in detecting anomalies within network traffic. Here they study combines Wireshark for packet filtering with TShark and Python for automating the process. The research also explores integrating machine learning models to further analyze traffic anomalies and improve detection rates. They have propose developing real-time monitoring systems using scripting techniques to handle various network conditions and threats.

#### **PROPOSED METHODOLOGY:**

**Data Collection:** Capture real-time network traffic using Wireshark, storing it in PCAP files. This traffic will contain both normal and potentially harmful packets.

**Data Filtering:** Leverage Wireshark's advanced filtering capabilities to detect anomalies in the network, such as malware, suspicious behaviors, or slow connections.

**Automated Traffic Capture:** Use TShark to automate the traffic capture process, enabling continuous monitoring of network activity.

**Geolocation Mapping:** Utilize Python, along with libraries like Pyshark and Geopy, to analyze the PCAP data and visually map the destination IP addresses.

**Testing and Validation:** Perform testing across different network environments to assess the effectiveness of the methodology in detecting network issues and malicious activities.

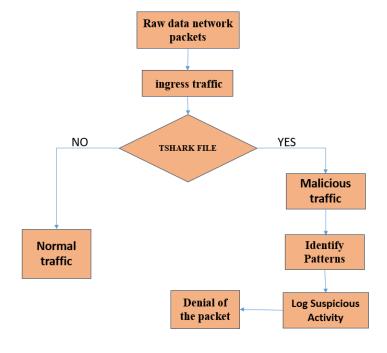


Fig 1 (WORKFLOW)

#### **IMPLEMENTATION:**

- 1. **Setup Environment:** Install Wireshark and Tshark on your Linux machine, ensuring you have the necessary permissions to capture network traffic.
- 2. **Monitoring network traffic**: it provides detailed information such as timestamps, frame numbers, and packet sizes. At the network layer, it reveals source and destination MAC addresses, IP addresses, and protocols (like TCP, UDP, or ICMP). You can also see transport layer details like port numbers, packet flags, sequence numbers, and payload data, helping analyze network communication and detect issues like latency, dropped packets, or security threats
- 3. Running Malware: DDoS and ARP Poisoning: When malware initiates a DDoS (Distributed Denial of Service) attack, Wireshark captures high volumes of traffic targeting specific IP addresses, often overwhelming a server with numerous requests. In ARP poisoning, Wireshark detects manipulated ARP packets where the malware alters MAC-to-IP mappings, redirecting network traffic. These attacks lead to potential denial of service or man-in-the-middle scenarios, revealing malicious intent in network activity logs.
- 4. **Capturing malware traffic:** When running malware in a controlled environment and capturing the network traffic with Wireshark, you can observe malicious activities like unusual connections, abnormal traffic patterns, or data exfiltration attempts.
- **5. Automating Network Analysis with Tshark:**By automating Tshark, you can efficiently scan live network traffic and analyze PCAP files. Tshark's filters allow you to capture specific packets or protocols in real-time, and automation scripts can quickly process and extract key information, such as IP addresses or unusual traffic patterns, from stored capture files. This makes network monitoring and malware detection more streamlined and less reliant on manual analysis.

#### **RESULT: OUTPUT**

#### NORMAL TRAFFIC CATURE

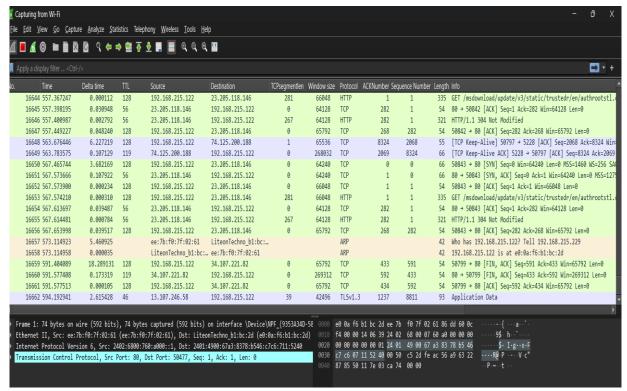


Fig 2(Normal Traffic)

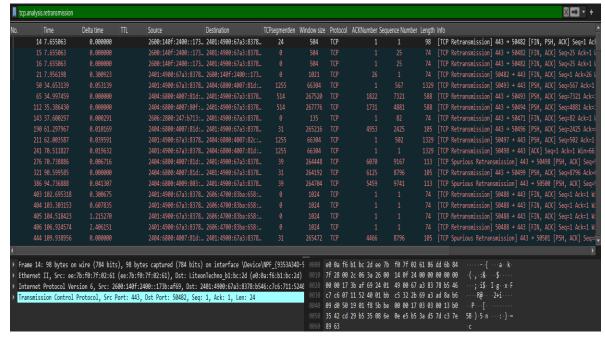


Fig 3(Retransmission issue)

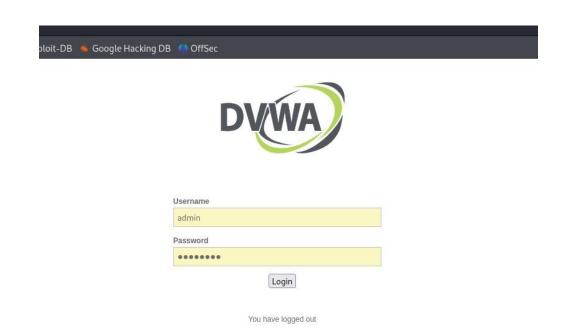
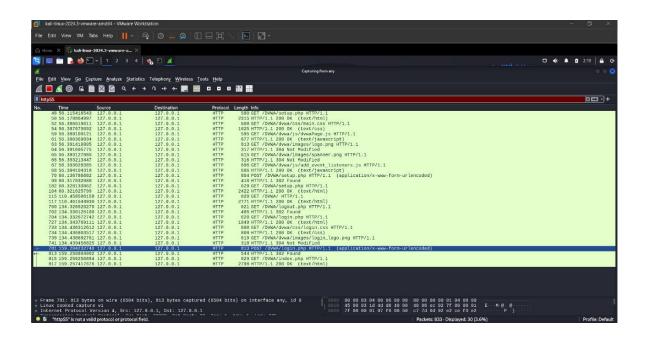


Fig 4



**Fig 5 (Scanning HTTP Traffic)** 

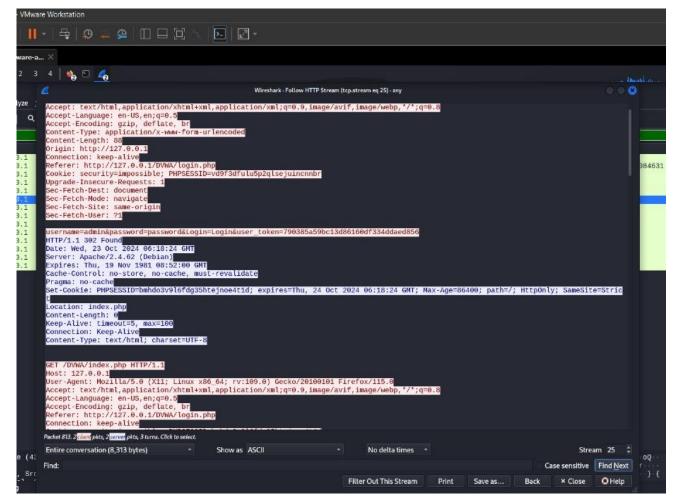


Fig 6 (Analyzing HTTP traffic)

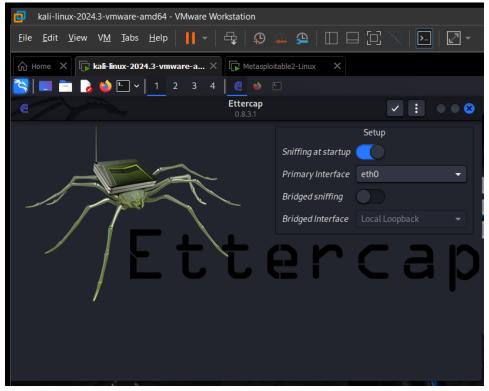


Fig 7 (Arp poisioning)

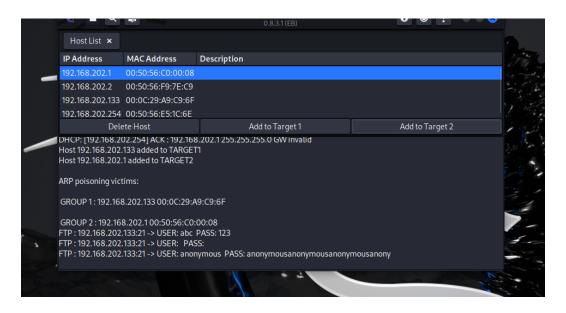


Fig 8 (HOST IP ADDRESS)

```
220 (vsFTPd 2.3.4)

OPTS UTF8 ON

200 Always in UTF8 mode.

USER anonymous

331 Please specify the password.

PASS anonymousanonymousanony

230 Login successful.
```

Fig 9 (Scanning Arp poisioning)

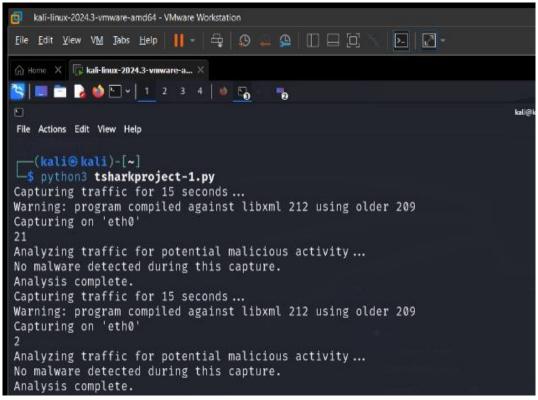


Fig 10 (Tshark Network Scaning)

Fig 11 (Tshark automation)

```
File Actions Edit View Help
kali@kali: ~ × kali@kali: ~/captures ×
 GNU nano 8.1
Counting TCP, UDP, and HTTP packets...
TCP packets: 410925
UDP packets: 49
HTTP packets: 8
Detecting potential DDoS attacks (SYN flood)...
Potential SYN flood detected with 205369 SYN packets
Detecting DNS spoofing ...
No DNS spoofing detected.
Checking for malicious domains ...
Detecting ARP spoofing...
No ARP spoofing detected.
Calculating network speed ...
Average network speed: 130901.5195348077 bytes/sec
Calculating average RTT and TCP window size...
Average RTT: 0.000428603 seconds
Average TCP window size: 278.734
```

**Fig 12** 

#### **APPENDICES**

### **APPENDIX-1: CODE COMPILER**

#### **Networkscaning.py**

```
import subprocess
import time
from datetime import datetime
# Variables
INTERFACE = "eth0" # Replace with your network interface
CAPTURE_DURATION = 15 # Capture traffic for 60 seconds in each round
LOG_FILE = "/home/kali/captures/malicious_activity.log"
def analyze_traffic(pcap_file, log_file):
  malware\_detected = False
  print("Analyzing traffic for potential malicious activity...")
  # Detect potential SYN flood attack (multiple SYN requests with no ACK)
  syn_flood = subprocess.run(
     ["tshark", "-r", pcap_file, "-T", "fields", "-e", "ip.src",
     "-Y", "tcp.flags.syn == 1 and tcp.flags.ack == 0"],
     capture_output=True, text=True).stdout
  syn_flood_ips = set([line for line in syn_flood.splitlines()])
  if len(syn_flood_ips) > 100:
     malware\_detected = True
```

```
with open(log_file, 'a') as f:
       f.write(f"{datetime.now()}: MALWARE DETECTED - SYN Flood Attack from
IP(s): {', '.join(syn_flood_ips)}\n")
     print(f"MALWARE DETECTED: SYN Flood Attack from IP(s): {',
'.join(syn_flood_ips)}")
  # Detect port scan (multiple connections to different ports from a single IP)
  port_scan = subprocess.run(
    ["tshark", "-r", pcap_file, "-T", "fields", "-e", "ip.src", "-e", "tcp.dstport",
     "-Y", "tcp.flags.syn == 1"],
     capture_output=True, text=True).stdout
  port_scan_ips = set([line.split()[0] for line in port_scan.splitlines() if len(line.split()) >
01
  if len(port_scan_ips) > 10:
    malware_detected = True
     with open(log_file, 'a') as f:
       f.write(f"{datetime.now()}: MALWARE DETECTED - Port Scan Attack from
IP(s): {', '.join(port_scan_ips)}\n")
    print(f"MALWARE DETECTED: Port Scan Attack from IP(s): {',
'.join(port_scan_ips)}")
  # Detect brute-force attempts (multiple failed login attempts, e.g., SSH)
  brute force = subprocess.run(
    ["tshark", "-r", pcap_file, "-T", "fields", "-e", "ip.src", "-Y", "ssh and (frame contains
'failure')"],
     capture_output=True, text=True).stdout
  brute_force_ips = set([line for line in brute_force.splitlines()])
  if len(brute_force_ips) > 10:
    malware_detected = True
     with open(log_file, 'a') as f:
       f.write(f"{datetime.now()}: MALWARE DETECTED - SSH Brute Force Attack
from IP(s): {', '.join(brute_force_ips)}\n")
```

```
print(f"MALWARE DETECTED: SSH Brute Force Attack from IP(s): {',
'.join(brute_force_ips)}")
  if not malware_detected:
    print("No malware detected during this capture.")
    with open(log_file, 'a') as f:
       f.write(f"{datetime.now()}: No malware detected during this capture.\n")
  else:
    print(f"Malware detected. Details logged in {log_file}")
  print("Analysis complete.")
def main():
  while True:
    print(f"Capturing traffic for {CAPTURE_DURATION} seconds...")
     capture_file =
f"/home/kali/captures/capture_{datetime.now().strftime('%Y%m%d%H%M%S')}.pcap"
     subprocess.run(["tshark", "-i", INTERFACE, "-a",
f"duration:{CAPTURE_DURATION}", "-w", capture_file])
     analyze_traffic(capture_file, LOG_FILE)
    time.sleep(10)
if __name__ == "__main__":
  main()
Pcapscan.py
```

import subprocess import statistics

```
import os
from datetime import datetime
# Define constants
CAPTURE_FILE = "/home/kali/Desktop/normaltrafficpcap.pcapng"
LOG_FILE = "/home/kali/captures/analysis_report3.log"
MALICIOUS_DOMAINS = ["malicious.com", "badactor.net"]
# Helper function to run TShark command
def run_tshark(command):
  result = subprocess.run(command, shell=True, stdout=subprocess.PIPE,
stderr=subprocess.PIPE)
  return result.stdout.decode('utf-8')
# Analyze normal traffic
def analyze_normal_traffic(pcap_file, log_file):
  with open(log_file, 'a') as log:
     log.write("Analyzing normal traffic...\n")
     # Source-Destination Flow Analysis
     flows = run_tshark(f"tshark -r {pcap_file} -T fields -e ip.src -e ip.dst | sort | uniq -c")
     log.write("Source-Destination Flows:\n" + flows + "\n")
     # Count of TCP, UDP, HTTP packets
     tcp_count = run_tshark(f"tshark -r {pcap_file} -Y 'tcp' | wc -l").strip()
     udp_count = run_tshark(f"tshark -r {pcap_file} -Y 'udp' | wc -l").strip()
     http_count = run_tshark(f"tshark -r {pcap_file} -Y 'http' | wc -l").strip()
     log.write(f"TCP Packet Count: {tcp_count}\n")
     log.write(f"UDP Packet Count: {udp_count}\n")
     log.write(f"HTTP Packet Count: {http_count}\n")
```

```
# Extract TCP Delta Time and Window Size
    delta_times = run_tshark(f"tshark -r {pcap_file} -T fields -e tcp.time_delta -Y
'tcp.time_delta'").splitlines()
    window_sizes = run_tshark(f"tshark -r {pcap_file} -T fields -e tcp.window_size -Y
'tcp.window_size'").splitlines()
    if delta_times:
       avg_delta_time = statistics.mean(map(float, delta_times))
    else:
       avg_delta_time = 0
    if window_sizes:
       avg_window_size = statistics.mean(map(int, window_sizes))
    else:
       avg\_window\_size = 0
    log.write(f"Average TCP Delta Time: {avg_delta_time:.6f} seconds\n")
    log.write(f"Average TCP Window Size: {avg_window_size:.2f} bytes\n")
    # Detect retransmissions
    retransmission_count = run_tshark(f"tshark -r {pcap_file} -Y
'tcp.analysis.retransmission' | wc -1").strip()
    log.write(f"TCP Retransmissions: {retransmission_count}\n")
    # Packet Loss Detection
    packet_loss_count = run_tshark(f"tshark -r {pcap_file} -Y 'tcp.analysis.lost_segment'
| wc -1").strip()
    log.write(f"Packet Loss Events: {packet_loss_count}\n")
# Detect DDoS attack
def detect_ddos(pcap_file, log_file):
```

```
with open(log_file, 'a') as log:
     log.write("Checking for potential DDoS attack...\n")
     ddos_detection = run_tshark(f"tshark -r {pcap_file} -T fields -e ip.src -e ip.dst -Y
'tcp.flags.syn == 1 and tcp.flags.ack == 0' | sort | uniq -c | awk '$1 > 100''')
     if ddos detection:
       log.write(f"DDoS Attack Detected:\n{ddos_detection}\n")
     else:
       log.write("No DDoS attack detected.\n")
# Detect DNS spoofing
def detect_dns_spoofing(pcap_file, log_file):
  with open(log_file, 'a') as log:
     log.write("Checking for DNS spoofing...\n")
     dns_spoof = run_tshark(f"tshark -r {pcap_file} -Y 'dns.flags.response == 1 &&
dns.qry.name != dns.a' -T fields -e ip.src -e dns.qry.name -e dns.a")
     if dns_spoof:
       log.write(f"DNS Spoofing Detected:\n{dns_spoof}\n")
     else:
       log.write("No DNS spoofing detected.\n")
# Check for malicious domains
def check_malicious_domains(pcap_file, log_file):
  with open(log_file, 'a') as log:
     log.write("Checking for malicious domains...\n")
     for domain in MALICIOUS DOMAINS:
       malicious_check = run_tshark(f"tshark -r {pcap_file} -Y 'dns.qry.name contains
{domain}' -T fields -e dns.qry.name")
       if malicious_check:
          log.write(f"Potential malicious domain detected: {domain}\n")
# Detect ARP spoofing
def detect_arp_spoofing(pcap_file, log_file):
```

```
with open(log_file, 'a') as log:
     log.write("Checking for ARP spoofing...\n")
     arp_spoof = run_tshark(f"tshark -r {pcap_file} -Y 'arp.duplicate-address-detected'")
    if arp_spoof:
       log.write(f"ARP Spoofing Detected:\n{arp_spoof}\n")
     else:
       log.write("No ARP spoofing detected.\n")
# Calculate network speed
def calculate_network_speed(pcap_file, log_file):
  with open(log_file, 'a') as log:
     log.write("Calculating network speed...\n")
     network_speed = run_tshark(f"tshark -r {pcap_file} -T fields -e frame.time_relative -e
frame.len | awk 'BEGIN{{prev_time=0;total_bytes=0;}}
{{time=$1;bytes=$2;if(prev_time!=0) {{total_bytes+=bytes;}}} prev_time=time;}}
END{{speed=total_bytes/(prev_time); print speed;}}'")
     log.write(f"Network Speed: {network_speed} bytes/sec\n")
# Calculate RTT (Round Trip Time)
def calculate_rtt(pcap_file, log_file):
  with open(log_file, 'a') as log:
     log.write("Calculating Round Trip Time (RTT)...\n")
    rtt = run_tshark(f"tshark -r {pcap_file} -Y 'tcp.flags.ack == 1' -T fields -e ip.src -e
ip.dst -e tcp.time_delta | sort | uniq")
     log.write(f"RTT (in seconds) for TCP ACKs:\n{rtt}\n")
# Main function to analyze traffic from a pcap file
def analyze_pcap():
  print(f"Analyzing traffic from {CAPTURE_FILE}...")
  # Analyze normal traffic and detect malicious activity
  analyze_normal_traffic(CAPTURE_FILE, LOG_FILE)
```

```
detect_ddos(CAPTURE_FILE, LOG_FILE)
  detect_dns_spoofing(CAPTURE_FILE, LOG_FILE)
  check_malicious_domains(CAPTURE_FILE, LOG_FILE)
  detect_arp_spoofing(CAPTURE_FILE, LOG_FILE)
  calculate_network_speed(CAPTURE_FILE, LOG_FILE)
  calculate_rtt(CAPTURE_FILE, LOG_FILE)
  print(f"Analysis complete. Results logged in {LOG_FILE}")
if __name__ == "__main__":
  analyze_pcap()
Running malware:
   –(kali⊕kali)-[~]
$\sudo hping3 -S --flood -V -p 80 192.168.202.132
[sudo] password for kali:
using eth0, addr: 192.168.202.132, MTU: 1500
HPING 192.168.202.132 (eth0 192.168.202.132): S set, 40 headers + 0 data bytes
hping in flood mode, no replies will be shown
--- 192.168.202.132 hping statistic ---
224601 packets transmitted, 0 packets received, 100% packet loss
```

#### Conclusion

The project proved that it is possible to capture, analyze, and visual network traffic with Wireshark, Python, and TShark. By assisting the automated process of traffic capture while mapping geolocation information, a clear approach is provided that assists in identifying what is normal traffic and what is suspected to be malicious in nature. Since this system also incorporates the analysis of data in real-time, it helps in the monitoring of networks and quicker resolution of cybersecurity issues.

#### **Future Scope**

Machine learning models could be incorporated for real-time anomaly detection as well as automating the entire process for larger amount of data and more complex traffic patterns. Also, embedding more recent threat-detection methods could also supplement the system's capabilities in recognizing modern era cyberattacks.

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## **WORKLOG**

Day	Date	Task Done
Day 1	18/08/2024	Network traffic data collection with Wireshark
Day 2	20/08/2024	Filtering and analyzing suspicious traffic patterns
Day 3	27/08/2024	Storing captured data in PCAP files
Day 4	30/08/2024	Python script development for geolocation mapping
Day 5	03/09/2024	Geolocation visualization using pyshark
Day 6	06/09/2024	Automating traffic capture using TShark
Day 7	12/09/2024	Debugging TShark automation
Day 8	16/09/2024	Testing automated capture in various network scenarios
Day 9	19/09/2024	Enhancing Python scripts to handle large datasets
Day 10	24/09/2024	Integrating geolocation with real-time data capture
Day 11	26/09/2024	Implementing error handling for traffic anomalies
Day 12	28/09/2024	Refining traffic filtering in Wireshark
Day 13	30/09/2024	Running extensive tests on different network protocols.
Day 14	01/10/2024	Data visualization of geolocations for normal traffic
Day 15	05/10/2024	Visualization of malicious traffic patterns on the map
Day 16	12/10/2024	Documenting the analysis process and findings
Day 17	19/10/2024	Final project testing and validation
Day 18	20/10/2024	Preparing the final report and visual aids
Day 19	22/10/2024	Uploading tasks and submission.