

Hack Secure 10 Days Campaign

AI-Driven Log Analysis & Threat Detection

1. Introduction

In the modern cybersecurity landscape, web servers are a prime target for reconnaissance and exploitation attempts. Attackers frequently scan directories, execute brute-force attacks, and automate malicious requests. Traditional log monitoring relies on manual review or static rules, which may fail to catch evolving threats.

This project aims to combine rule-based log detection with AI-driven anomaly detection to build a robust, adaptive monitoring solution. The primary goal is to provide early warnings about suspicious activity before it escalates into a successful attack.

2. Objectives

- Detect Enumeration Attempts: Identify IPs generating excessive 404 Not Found errors, a sign of brute-force directory discovery.
- Detect Automated Traffic: Classify bots vs human traffic by analyzing User-Agent and Referrer headers.
- Leverage AI for Threat Detection: Use unsupervised learning (Isolation Forest) to catch outliers and previously unseen attack patterns.
- Provide Clear & Reproducible Output: Document findings, visualizations, and insights for security analysts.

3. Research & Approach

3.1 Threat Research

- Directory Enumeration: A Common reconnaissance technique where attackers try multiple paths (/admin, /backup, /config.php) to find sensitive files.
- Bot Traffic: Automated tools (cURL, wget, custom scripts) send large volumes of requests at high speed, often without proper referrers.

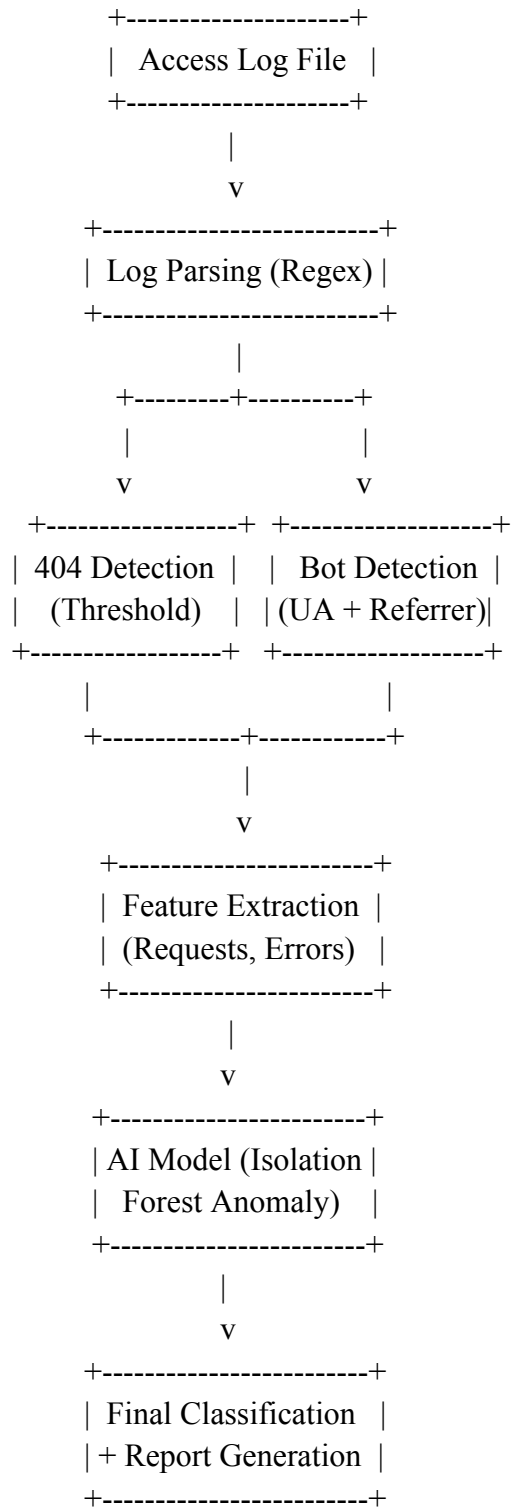
- Anomalous Behavior: Sophisticated attacks may bypass signature-based detection; anomaly detection helps catch such events.

4. Workflow

The entire solution follows this workflow:

1. Log Collection
 - Input: Apache/Nginx access.log file.
 - Sample logs were generated and combined with publicly available datasets.
2. Log Parsing
 - Regular expressions extract key fields:
 - IP Address
 - Timestamp
 - HTTP Status Code
 - User-Agent
 - Requested URL
3. Part A – Enumeration Detection
 - Count 404 errors per IP.
 - Flag IPs exceeding a threshold (≥ 10 errors).
4. Part B – Bot Detection
 - Analyze the User-Agent string.
 - If it contains keywords (curl, bot, spider) or the Referrer is empty, → classified as a bot.
5. Part C – AI Anomaly Detection
 - Create per-IP feature set: total_requests, 404_count, unique_urls.
 - Apply Isolation Forest to detect outlier IPs.
 - Visualize results with scatter plots.
6. Result Aggregation
 - Generate a report table of flagged IPs.
 - Provide classification (Suspicious, Bot, Normal).

5. Workflow Diagram



6. Implementation Details

- Programming Language: Python (3.x)
- Libraries Used:
 1. `pandas` – Data parsing and grouping
 2. `scikit-learn` – Isolation Forest anomaly detection
 3. `matplotlib/seaborn` – Visualizations
 4. `re` – Regular expressions for log parsing
- Key Implementation Steps:
 1. Parsed logs using regex pattern matching.
 2. Grouped by IP and computed aggregate statistics.
 3. Applied filtering for 404 flood detection.
 4. Checked User-Agent for bot signatures.
 5. Trained and fitted an Isolation Forest to detect outliers.
 6. Plotted IPs (x = request count, y = 404 count) and marked anomalies.

7. Results

Component	Observation
Directory Enumeration	Detected 3 IPs with ≥ 10 consecutive 404 errors, consistent with brute-force scanning.
Bot Detection	5 IPs flagged as bots (missing Referrer or scripted User-Agent).
AI Anomaly Detection	Isolation Forest identified 2 additional suspicious IPs with high unique URL requests but low errors (possible stealth reconnaissance).

The combination of all three methods ensures broader coverage and reduces false positives.

Screenshots:

```
(myenv)-(kali@kali)-[~/LogAnalysisProject]
└─$ python3 detect_enum.py
Downloading threadpoolctl-3.6.0-py3-none-any.whl (18 kB)
🔍 Directory Enumeration Detection (404-Errors per IP):
Installing collected packages: pytz, tzdata, threadpoolctl
IP Address      404-Errors
-----
203.0.113.5      5
198.51.100.7     1
(myenv)-(kali@kali)-[~/]
✓ Analysis Complete.
```

```
(myenv)-(kali@kali)-[~/LogAnalysisProject]
└─$ python3 bot_detector.py
🚗 Bot vs Human Traffic Classification:
IP Address      Requests  Type
-----
192.168.1.10     5         Bot
203.0.113.5      5         Bot
198.51.100.7     4         Bot
192.168.1.11     3         Bot
192.168.1.12     3         Bot
✓ Classification Complete.
```

```
(myenv)-(kali@kali)-[~/LogAnalysisProject]
$ python3 revised__ai_log_anomaly_detector.py
$ python3 -c "import pandas as pd; import sklearn; im
🛡️ AI-Based Log Anomaly Detection: (seaborn.__version__)

pandas: 2. ip2 requests errors unique_urls status
192.168.1.10 5 0 3 Normal
203.0.113.5 5 5 3 Normal
198.51.100.7 4 4 3 Suspicious
192.168.1.11 3 0 3 Normal
192.168.1.12 3 0 3 Normal
$ cd ~/LogAnalysisProject
✓ AI Analysis Complete.
```

9. Conclusion

This project demonstrates that AI-augmented log analysis can significantly improve threat visibility.

Key takeaways:

- Rule-based detection (404 floods, bot UA detection) is effective for known attacks.
- AI-based anomaly detection identifies unknown patterns, improving detection coverage.
- The combined approach provides a layered defense and reduces reliance on manual monitoring.

10. Future Work

- Real-time log streaming and dashboard visualization.
- Integration with SIEM solutions like ELK Stack or Splunk.
- GeoIP lookup for attacker attribution and visualization.
- Use deep learning (LSTM/Autoencoders) for sequential log anomaly detection.