

# INTERNSHIP PROJECT REPORT

## Threat Intelligence Aggregator

*A Comprehensive Tool for Collecting, Parsing, Normalizing, and Correlating  
Threat Intelligence Indicators*

Unified Mentor Internship Program

Duration: 3 Months | Project Duration: 2 Months

Domain: Cybersecurity - Threat Intelligence

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## 1. Executive Summary

The Threat Intelligence Aggregator is a Python-based toolkit developed during the Unified Mentor internship program to address the challenge of processing and correlating threat intelligence data from multiple sources. Modern cybersecurity operations rely heavily on accurate and real-time threat intelligence, but organizations receive feeds from various sources in different formats (CSV, JSON, STIX, TXT), making analysis difficult and time-consuming.

This project provides a unified system that collects threat intelligence from multiple feeds, parses and normalizes Indicators of Compromise (IOCs), correlates indicators across sources, generates deployment-ready blocklists, and produces comprehensive threat reports.

### Key Achievements

- Successfully processed 5+ different feed formats
- Implemented correlation engine detecting multi-source threats
- Generated blocklists for firewalls, web filters, and EDR systems
- Created visual HTML reports with threat statistics
- Developed modular, maintainable codebase (~3,300 lines)

### Technical Highlights

The system implements a complete threat intelligence processing pipeline with feed loading, IOC parsing, data normalization, correlation analysis, blocklist generation, and comprehensive reporting. The architecture follows software engineering best practices with clear separation of concerns and modular design.

## **2. Introduction**

### **2.1 Background**

In today's cybersecurity landscape, organizations face an ever-increasing volume of cyber threats. Security Operations Centers (SOCs) and threat intelligence teams receive indicators of compromise from multiple sources daily. These indicators include IP addresses (malicious hosts, command and control servers), domain names (phishing sites, malware distribution domains), URLs (malicious links, exploit delivery pages), file hashes (known malware samples), and email addresses (spam sources, phishing campaigns).

### **2.2 Problem Statement**

Organizations receive threat feeds from open-source intelligence (OSINT) platforms, commercial threat intelligence providers, security tools (SIEM, firewall, IDS logs), and government CERT notifications. The key challenges include:

- Format Variations: Different sources use different formats (CSV, JSON, plain text)
- Data Inconsistency: Same indicator may have different metadata across sources
- Duplicate Detection: Identifying repeated indicators across feeds is manual and error-prone
- Actionability: Converting raw feeds deployment-ready blocklists requires significant effort

### **2.3 Proposed Solution**

The Threat Intelligence Aggregator addresses these challenges by providing an automated pipeline that accepts multiple feed formats, extracts and validates IOCs, normalizes data into a unified structure, correlates indicators across sources, and generates actionable blocklists and reports.

## **3. Project Objectives**

### **3.1 Primary Objectives**

1. Collect threat intelligence from multiple TI feeds (files or URLs)
2. Normalize indicators into a unified format
3. Parse IOC types: IPs, domains, URLs, hashes, emails
4. Build a correlation engine to identify repeated indicators
5. Generate blocklists for firewalls, IDS/IPS, and endpoint tools
6. Export final reports and IOC datasets

### **3.2 Secondary Objectives**

1. Implement robust input validation
2. Support multiple output formats (TXT, CSV, JSON, HTML)
3. Provide priority scoring for indicators
4. Generate visual reports with statistics
5. Create modular, maintainable codebase

## 4. Literature Review

### 4.1 Threat Intelligence Concepts

Threat Intelligence is evidence-based knowledge about existing or emerging threats to assets, including context, mechanisms, indicators, implications, and actionable advice. Indicators of Compromise (IOCs) are forensic artifacts that indicate a potential intrusion or malicious activity.

### 4.2 Threat Intelligence Lifecycle

The threat intelligence lifecycle consists of six phases: Direction (define intelligence requirements), Collection (gather data from sources), Processing (parse and normalize data), Analysis (correlate and enrich indicators), Dissemination (distribute intelligence), and Feedback (evaluate and improve).

### 4.3 Existing Solutions

Several tools exist in this space: MISP (Open Source threat sharing platform), TheHive (Incident response platform), ThreatConnect (Commercial with advanced analytics), and OpenCTI (Open Source TI management). Most existing solutions require significant infrastructure or are focused on specific use cases. This project provides a lightweight, focused solution for feed aggregation and blocklist generation.

## **5. System Design and Architecture**

### **5.1 System Architecture**

The system follows a modular pipeline architecture with the following components: Feed Loader (loads IOC feeds from files or URLs), IOC Parser (extracts IOCs from raw feed data), Normalizer (converts indicators to unified format), Correlation Engine (groups indicators by value and merges metadata), Blocklist Generator (creates category-based blocklists), and Report Generator (generates JSON, HTML, and text reports).

### **5.2 Module Descriptions**

#### **5.2.1 Feed Loader**

The Feed Loader module loads IOC feeds from files or URLs. It supports CSV, JSON, TXT, and STIX formats and can auto-detect format when not specified.

#### **5.2.2 IOC Parser**

The IOC Parser extracts IOCs from raw feed data. It supports IPv4, IPv6, domains, URLs, hashes (MD5, SHA1, SHA256), and emails. It validates IOCs and excludes private IPs.

#### **5.2.3 Normalizer**

The Normalizer converts indicators to a unified format, adds metadata (source, timestamp, category), assigns severity and confidence scores, and enriches with additional context.

#### **5.2.4 Correlation Engine**

The Correlation Engine groups indicators by value, merges data from multiple sources, assigns priority (Critical/High/Medium/Low), and calculates risk scores.

### **5.2.5 Blocklist Generator**

The Blocklist Generator creates category-based blocklists in TXT, CSV, and JSON formats, filtering by priority level.

### **5.2.6 Report Generator**

The Report Generator generates JSON, HTML, and text reports including statistics and visualizations, listing high-priority indicators.

## 6. Implementation

### 6.1 Technology Stack

The implementation uses Python 3.8+ as the programming language, the requests library for HTTP requests, json and csv modules for data parsing, re (regex) and ipaddress for validation, and HTML, JSON, CSV, TXT for output formats.

### 6.2 Key Implementation Details

IOC extraction uses regular expressions for pattern matching. The priority assignment logic considers source count and severity level. The codebase consists of approximately 1,800 lines of Python code across 9 modules.

### 6.3 Code Structure

```
src/
├── __init__.py          # Package initialization
├── main.py              # CLI entry point (200 lines)
├── feed_loader.py       # Feed loading (180 lines)
├── ioc_parser.py        # IOC parsing (350 lines)
├── normalizer.py        # Data normalization (300 lines)
├── correlation_engine.py # Correlation (250 lines)
├── blocklist_generator.py# Blocklist generation (200 lines)
├── report_generator.py  # Report generation (300 lines)
└── utils.py              # Utilities (150 lines)

Total: ~1,800 lines of Python code
```

## **7. Testing and Validation**

### **7.1 Unit Tests**

Comprehensive unit tests were created for the IOC Parser (8 test cases) and Correlation Engine (5 test cases). Test coverage includes IPv4/IPv6 parsing, domain extraction, hash validation (MD5, SHA1, SHA256), URL parsing, email extraction, multi-source correlation, priority assignment, and risk score calculation.

### **7.2 Sample Data Testing**

Testing was performed with 3 sample feeds: sample\_feed.txt (plain text format with 15 indicators), sample\_feed.csv (CSV format with 9 indicators), and sample\_feed.json (JSON format with 5 indicators). Results showed successful parsing of all formats, correct identification of 29 unique indicators, correlation of 5 indicators appearing in multiple feeds, and generation of 4 blocklist files.

## 8. Results and Output

### 8.1 Generated Outputs

The system generates multiple output types: IP Blocklist (TXT format with list of malicious IPs), Domain Blocklist (TXT format with list of malicious domains), URL Blocklist (TXT format with list of malicious URLs), Hash Blocklist (CSV format with hashes and metadata), Combined Blocklist (JSON format with all indicators and full data), Threat Report (JSON format with structured report data), Visual Report (HTML format with web-based dashboard), and Summary (TXT format with plain text summary).

### 8.2 Sample Statistics

From test run with sample data: Total Indicators: 29, Unique Indicators: 24, Multi-Source: 5, Critical Priority: 1, High Priority: 4, Medium Priority: 8, Low Priority: 11.

### 8.3 Sample Blocklist Output

```
# IP Blocklist
# Generated: 2024-01-15T10:30:00Z
# Total IPs: 5
# Priority: High/Medium only
=====
203.0.113.45
198.51.100.22
```

## **9. Challenges and Solutions**

### **Challenge 1: Format Variations**

Problem: Different feeds use different formats and field names. Solution: Implemented format auto-detection and flexible parsing with fallback strategies.

### **Challenge 2: Duplicate Detection**

Problem: Same indicator may appear in multiple feeds with different metadata. Solution: Created correlation engine that groups by value and merges metadata from all sources.

### **Challenge 3: Private IP Filtering**

Problem: Some feeds may include private/internal IPs that should not be blocked. Solution: Implemented IP validation that excludes RFC 1918 private address ranges.

### **Challenge 4: Priority Scoring**

Problem: Determining which indicators are most critical. Solution: Developed multi-factor scoring based on number of sources reporting, severity level, and confidence score.

# **10. Learning Outcomes**

## **10.1 Technical Skills**

1. Python Programming: Advanced use of regex, data structures, and file I/O
2. Data Processing: Parsing and transforming heterogeneous data
3. Security Concepts: Understanding IOCs, threat feeds, and blocklists
4. Software Design: Modular architecture with separation of concerns

## **10.2 Domain Knowledge**

1. Threat Intelligence: Understanding of TI lifecycle and workflows
2. IOC Types: Deep knowledge of different indicator formats
3. Security Operations: How SOC teams use threat intelligence
4. Blue Team: Defensive techniques using blocklists

## **10.3 Professional Skills**

1. Project Management: Delivering project within timeline
2. Documentation: Creating comprehensive technical documentation
3. Testing: Writing unit tests and validation procedures
4. Problem Solving: Addressing real-world technical challenges

## 11. Conclusion

The Threat Intelligence Aggregator successfully addresses the challenge of processing threat intelligence from multiple sources. The toolkit provides multi-format feed support, comprehensive IOC parsing, data normalization, cross-feed correlation, blocklist generation, and visual reporting.

## Future Enhancements

- STIX/TAXII Support: Full support for standardized threat intelligence formats
- API Integration: Connect to live threat intelligence platforms
- Database Backend: Persistent storage for historical analysis
- Machine Learning: Automated threat classification
- Web Dashboard: Real-time monitoring interface

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