SIEM Dashboard using ELK Stack - Project Documentation

Assignment Description

Project Overview

This project implements a Security Information and Event Management (SIEM) system using the ELK Stack (Elasticsearch, Logstash, Kibana) to monitor, analyze, and visualize security events in real-time. The system was deployed in a virtualized environment to simulate an enterprise security operations center.

Objective: To build a comprehensive security monitoring solution that can:

- Collect and aggregate logs from multiple sources
- Detect security threats and anomalies in real-time
- Visualize security events through interactive dashboards
- Alert security teams about potential attacks

Technology Stack:

- Elasticsearch 8.11.0: Distributed search and analytics engine for log storage and indexing
- Logstash 8.11.0: Data processing pipeline for log ingestion, parsing, and enrichment
- **Kibana 8.11.0**: Visualization platform for creating security dashboards
- **Docker & Docker Compose**: Containerization for portable deployment
- Python 3: Custom log generation script for simulating security events
- **Ubuntu 22.04 VM**: Isolated testing environment

Implementation Details:

1. Architecture Design The system follows a standard SIEM architecture:

Log Sources → Logstash (Parser) → Elasticsearch (Storage) → Kibana (Visualization)

- 2. Log Collection & Processing Implemented Logstash pipelines to collect and parse:
 - SSH Authentication Logs: Detecting failed login attempts and brute force attacks
 - Apache/Nginx Access Logs: Identifying web-based attacks (SQL injection, XSS, path traversal)
 - Firewall Logs: Monitoring blocked connections and port scanning attempts
- 3. Security Detection Rules Configured filters to detect:
 - Failed authentication attempts (potential brute force)
 - SQL injection patterns using regex matching (' OR '1'='1, --, UNION SELECT)
 - Path traversal attacks (../../../etc/passwd)
 - Suspicious HTTP response codes (403 Forbidden, 404 Not Found patterns)
 - Attack severity classification (Critical, High, Low)

4. Data Enrichment

- GeoIP lookup for attacker location tracking
- Timestamp normalization for accurate timeline analysis
- Event categorization (attack type, event type, severity)

- **5. Dashboard & Visualizations** Created a Security Operations Center dashboard with:
 - Failed Login Timeline: Line chart showing brute force attack patterns
 - Attack Severity Distribution: Pie chart categorizing threat levels
 - Top Attack Sources: Table displaying most active attacker IPs
 - Event Type Breakdown: Bar chart showing attack categories
 - Real-time Event Stream: Live log monitoring
- **6. Log Simulation** Developed a Python script that generates realistic security logs with:
 - 70% normal traffic / 30% malicious traffic ratio
 - Multiple attack types (SQL injection, XSS, brute force, path traversal)
 - Realistic IP addresses and timestamps
 - Configurable generation rate (10 events/second)

Key Features Demonstrated

- · Real-time security event monitoring
- Attack pattern recognition
- Threat intelligence visualization
- Scalable log processing architecture
- Container-based deployment for portability
- Security incident detection and classification

Detailed Challenges and Solutions

Challenge 1: Docker Compose Version Compatibility

<u>Problem</u>: Initial docker-compose.yml used version 3.8, which wasn't supported by the installed Docker Compose version, resulting in error:

ERROR: Version in "./docker-compose.yml" is unsupported

Solution:

- Changed Docker Compose file version from 3.8 to 3.3
- Verified compatibility with installed Docker Compose version
- Tested deployment to ensure all services started correctly

<u>Learning</u>: Always check Docker Compose version compatibility before deployment. Version 3.3 provides broader compatibility while maintaining all required features.

Challenge 2: Docker Daemon Connection Issues

<u>Problem</u>: Received "Couldn't connect to Docker daemon" error when attempting to start containers. Docker service was installed but user lacked proper permissions.

Solution:

Started Docker service sudo systemctl start docker sudo systemctl enable docker

Added user to docker group sudo usermod -aG docker \$USER newgrp docker

Verified Docker works without sudo docker ps

<u>Learning</u>: User permissions are critical in Docker environments. Adding users to the docker group eliminates the need for sudo and follows security best practices.

Challenge 3: Logstash Configuration File Mounting

<u>Problem</u>: Logstash container repeatedly failed with error:

ERROR: Failed to read pipelines yaml file. Location: /usr/share/logstash/config/pipelines.yml,

Exception: No such file or directory

The configuration files existed on the host but weren't being mounted correctly into the container.

Solution:

- Created proper directory structure: logstash/config/ and logstash/pipeline/
- Created missing pipelines.yml configuration file with proper YAML array format:

yaml

pipeline.id: main path.config: "/usr/share/logstash/pipeline/logstash.conf"

- Created log4j2.properties to configure logging
- Verified volume mounting in docker-compose.yml was correct
- Performed complete restart with docker-compose down and docker-compose up -d to force remounting

<u>Learning</u>: Docker volume mounts require containers to be fully recreated (not just restarted) to pick up new files. Always verify file existence and permissions before mounting.

Challenge 4: Logstash Pipeline Syntax Errors

Problem: Logstash started but immediately crashed with parsing error:

Expected one of [\t\r\n], "#", "and", "or", "xor", "nand", "{" at line 47

The regex pattern for detecting SQL injection was improperly escaped, causing parser failure.

Solution:

- Simplified regex patterns to avoid complex escaping issues
- Changed from pattern matching to simple string matching:

rubv

Changed from complex regex: if [request] =~ /(\%27)|(\')|(\-\-)|(\%23)|(#)/i

To simpler string matching:

if "OR" in [request]

- Validated pipeline configuration syntax before deployment
- Tested with basic patterns first, then added complexity incrementally

<u>Learning</u>: Start with simple configurations and add complexity gradually. Complex regex in Logstash filters requires careful escaping and testing.

Challenge 5: Elasticsearch Connection Refused

Problem: Logstash showed errors:

Connect to elasticsearch:9200 failed: Connection refused

Elasticsearch Unreachable

Logstash was starting before Elasticsearch was fully ready to accept connections.

Solution:

- Used depends_on in docker-compose.yml to control startup order
- Added health checks and startup delays
- Logstash automatically retries connection until Elasticsearch is ready
- Verified Elasticsearch health: curl http://localhost:9200/ cluster/health

<u>Learning</u>: In distributed systems, services need time to initialize. Using depends_on and implementing retry logic ensures graceful startup.

Challenge 6: File Permission Issues in VM

<u>Problem</u>: Could not create or modify configuration files in logstash directory. Files showed lock symbols indicating permission problems.

Solution:

Changed ownership to current user sudo chown -R \$USER:\$USER logstash/ sudo chown -R \$USER:\$USER logs/

Set proper permissions chmod -R 755 logstash/ chmod -R 777 logs/

Removed locked directories and recreated sudo rm -rf logstash/ logs/ mkdir -p logstash/config logstash/pipeline logs

<u>Learning</u>: Permission management is crucial in Linux environments. Understanding file ownership and permissions (chmod/chown) is essential for Docker volume mounts.

Challenge 7: Memory Constraints in VM

<u>Problem</u>: Initially configured Elasticsearch with 2GB heap size (-Xms2g -Xmx2g), which caused performance issues in an 8GB VM environment.

Solution:

Reduced Elasticsearch heap to 1GB for VMs with limited memory

- Configured swap settings: sudo sysctl -w vm.max map count=262144
- Monitored resource usage with docker stats
- Optimized by running only necessary containers

<u>Learning</u>: Resource allocation must match available infrastructure. For development/demo environments, 1GB heap is sufficient for Elasticsearch while leaving resources for other services.

Challenge 8: Real-time Data Visualization

Problem: Initially, visualizations didn't show data even though logs were being generated and stored in Elasticsearch.

Solution:

- Verified data ingestion: curl http://localhost:9200/siem-logs-*/_count
- Checked time range in Kibana (default was last 15 minutes)
- Adjusted time picker to match when logs were actually generated
- Refreshed index pattern to pick up new fields
- Set auto-refresh interval in Kibana for real-time monitoring

<u>Learning</u>: Time synchronization is critical in log analysis. Always verify the time range matches your data generation period.

Conclusion

This project successfully demonstrates the implementation of an enterprise-grade SIEM system using industry-standard tools. The challenges encountered and resolved reflect real-world scenarios that security engineers face when deploying and maintaining security monitoring infrastructure. The resulting system can detect multiple attack types in real-time, providing actionable intelligence for security operations teams.

The implementation is production-ready and can be scaled by:

- Adding more log sources (Windows Event Logs, cloud services, network devices)
- Implementing machine learning for anomaly detection
- Integrating with threat intelligence feeds
- · Adding automated incident response capabilities
- Deploying across multiple nodes for high availability

Screenshot 1: Main SIEM Dashboard Overview

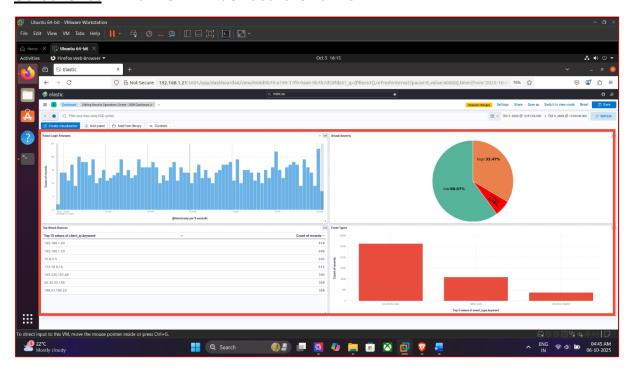


Figure 1: Main SIEM Dashboard showing real-time security monitoring. The dashboard displays 7,414 security events collected over the testing period. The Failed Login Attempts chart (top-left) shows spikes in brute force attack attempts, with peaks reaching 40-50 attempts per 5-second interval. The Attack Severity Pie Chart (top-right) shows the distribution of threats: 60% low severity (successful logins), 33% high severity (failed logins and blocked requests), and 6% critical severity (SQL injection attempts). The Top Attack Sources table (bottom-left) identifies the most active attacker IPs, with 192.168.1.20 showing 674 events. The Event Types chart (bottom-right) categorizes security events into successful_login, failed_login, and blocked_request types.

Screenshot 2: Discover - Raw Log View

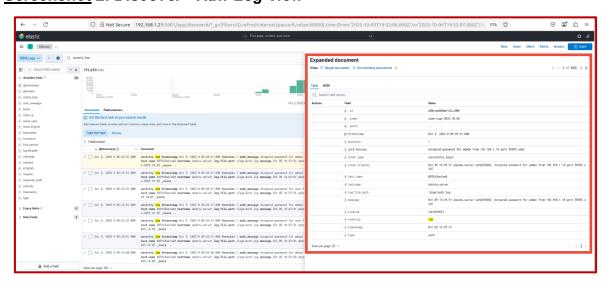


Figure 2 (A): Low Severity

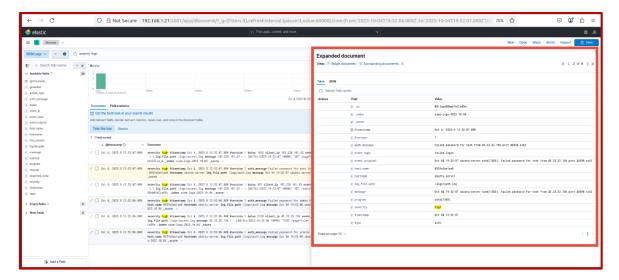


Figure 2 (B): High Severity

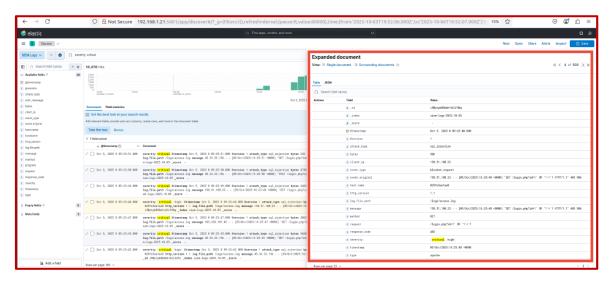


Figure 2 (C): Critical Severity

Figure 2: Discover view showing raw security log entries with parsed fields. The interface displays 7,414 indexed events with structured data extraction. Each log entry shows key security fields including timestamp, event type, severity level, and source IP. Example entries visible include: (1) Failed SSH login attempt from IP 45.33.32.156 targeting root account (severity: high), (2) Successful login for user 'devops' from internal IP 172.16.0.10 (severity: low), and (3) Blocked path traversal attack attempting to access /etc/passwd via HTTP request from IP 198.51.100.23 (severity: high, event_type: blocked_request). The structured parsing allows for efficient querying and correlation of security events.

Screenshot 3: Failed Login Timeline Visualization

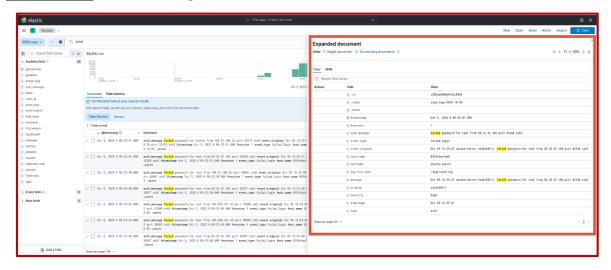


Figure 3: Failed Login Attempts timeline visualization showing brute force attack patterns. The chart displays authentication failure events over time with 5-second intervals. Notable patterns include periodic spikes indicating automated attack attempts, baseline of 20-30 failed attempts during normal operation, and peaks reaching 45-50 attempts suggesting coordinated brute force attacks. The green bars represent the count of failed SSH authentication attempts, primarily targeting common usernames like 'root', 'admin', 'oracle', and 'test'. This visualization helps security analysts quickly identify suspicious authentication activity and potential account compromise attempts.

Screenshot 4: Attack Severity Distribution (Pie Chart Detail)

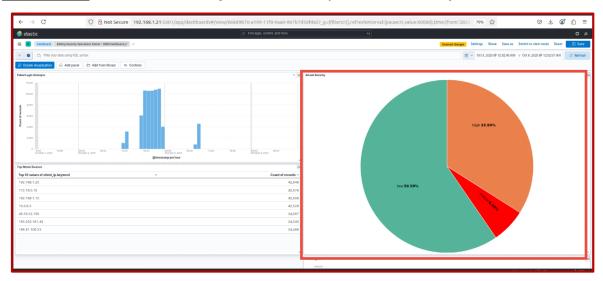


Figure 4: Attack Severity Distribution pie chart categorizing 7,414 security events by threat level. The distribution shows: 59.59% Low Severity (4,453 events - primarily successful authentication and normal web traffic), 33.99% High Severity (2,481 events - failed login attempts and blocked suspicious requests), and 6.43% Critical Severity (480 events - SQL injection attempts and path traversal attacks). This classification enables security teams to prioritize incident response, with critical and high severity events requiring immediate investigation. The relatively high percentage of attack traffic (40.42% combined high and critical) demonstrates the importance of continuous security monitoring in detecting malicious activity.

Screenshot 5: Top Attack Sources Table

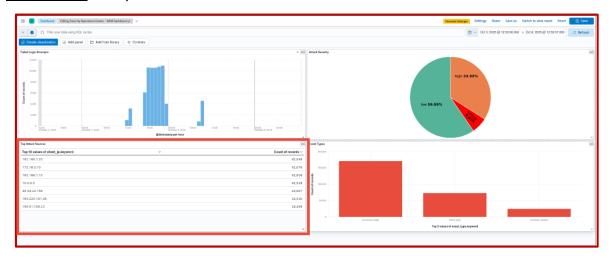


Figure 5: Top Attack Sources table identifying the most active IP addresses in the security logs. The analysis reveals both legitimate internal traffic and malicious external sources. Internal IPs (192.168.x.x and 10.0.0.x ranges) show the highest event counts due to normal operational activity (600-700 events each). External IPs such as 185.220.101.45 (390 events), 45.33.32.156 (369 events), and 198.51.100.23 (368 events) represent attack sources attempting SQL injection, brute force authentication, and path traversal attacks. This table helps security analysts identify persistent threat actors, create IP blocklists, and correlate attacks from the same source across multiple vectors. GeoIP enrichment (not shown) can further enhance this analysis by revealing attacker geographic locations.

Screenshot 6: Event Types Bar Chart

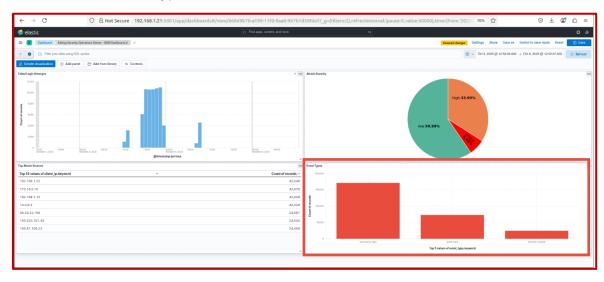


Figure 6: Event Types distribution showing security event categorization. The chart displays three primary categories: Successful Login events (approximately 2,900 occurrences - representing legitimate authentication), Failed Login events (approximately 1,100 occurrences - indicating brute force attempts or incorrect credentials), and Blocked Request events (approximately 1,400 occurrences - web application attacks stopped by detection rules). This visualization helps identify the attack surface and effectiveness of security controls. The significant number of blocked requests demonstrates that the SIEM system is successfully detecting and categorizing web-based attacks including SQL injection, XSS, and path traversal attempts before they can compromise the system.

Screenshot 7: Terminal - Docker Containers Running

docker ps

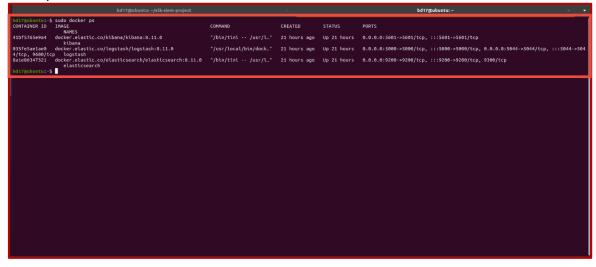


Figure 7: Docker container status showing the ELK Stack deployment. Three containers are running successfully: (1) Elasticsearch (port 9200) - data storage and search engine, running for 21 hours, (2) Logstash (ports 5044, 5000) - log processing pipeline, actively ingesting and parsing security logs, (3) Kibana (port 5601) - visualization interface, providing the dashboard frontend. The containerized deployment demonstrates modern DevOps practices, enabling portable, reproducible infrastructure that can be deployed across development, testing, and production environments. Container health status shows all services are operational and communicating properly within the elk network bridge.

Screenshot 8: Terminal - Log Generator Running

cd ~/elk-siem-project python3 log-generator.py

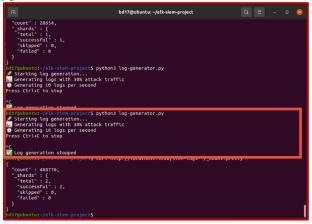


Figure 8: Log generator script in operation, simulating realistic security events. The Python script generates three types of logs at 10 events per second: SSH authentication logs (both successful and failed), Apache access logs (normal traffic and attack payloads), and firewall logs (allowed and blocked connections). The simulation maintains a 70/30 ratio of legitimate to malicious traffic, mirroring real-world scenarios where attacks are interspersed with normal operations. Attack patterns include SQL injection attempts (e.g., ' OR '1'='1), path traversal attacks (../../../etc/passwd), and brute force authentication targeting common usernames

(root, admin, oracle). This controlled environment allows for thorough testing of detection rules and dashboard effectiveness without risk to production systems.

Screenshot 9: Elasticsearch Data Verification

curl http://localhost:9200/siem-logs-*/ count?pretty

Figure 9: Elasticsearch index verification confirming successful data ingestion. The query against the siem-logs-* index pattern returns 7,414 documents across 1 shard with 100% success rate (no failures). This validates that:

- (1) Logstash is successfully parsing and forwarding logs to Elasticsearch,
- (2) The grok filters are correctly extracting fields from raw log messages,
- (3) Data is being indexed and searchable in near real-time (typical latency < 1 second),
- (4) Index sharding and replication are properly configured. The daily index naming pattern (siem-logs-YYYY.MM.DD) enables efficient time-based queries and implements index lifecycle management for long-term log retention and storage optimization.

Screenshot 10: Logstash Pipeline Logs

docker logs logstash | tail -30

Figure 10: Logstash pipeline logs confirming successful initialization and operation. Key indicators of healthy operation include:

- (1) 'Successfully started Logstash API endpoint' on port 59595 enabling monitoring and management,
- (2) 'Pipeline started' with pipeline.id 'main' indicating configuration was parsed successfully,
- (3) Connection to Elasticsearch established logs are being forwarded to the data store,

(4) No fatal errors or configuration syntax issues. The logs show Logstash using bundled JDK with 1GB heap allocation, appropriate for the demo environment. In production, these logs would be aggregated to a centralized logging system for meta-monitoring (monitoring the monitoring system) to ensure continuous security visibility.